

AN APPLICATION OF THE HOFFMAN-RIBAK METHOD

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ABSTRACT

We develop an algorithm for setting up initial Gaussian random density and velocity fields containing one or more peaks or dips, in an arbitrary cosmological scenario. The intention is to generate appropriate initial conditions for cosmological N-body simulations that focus on the evolution of the progenitors of the present-day galaxies and clusters. The procedure is an application of the direct and accurate prescription of Hoffman & Ribak (1991) for generating constrained random fields.

For each peak a total of 21 physical characteristics can be specified, including its scale, position, density Hessian, velocity, and velocity gradient. The velocity (or, equivalently, gravity) field constraints are based on a generalization of the formalism developed by Bardeen et al. (1986). The resulting density field is sculpted such that it induces the desired amount of net gravitational and tidal forces.

We provide a detailed mathematical presentation of the formalism. Afterwards we provide analytical estimates of the likelihood of the imposed constraints. Amongst others, it is shown that the tidal field has a strong tendency to align itself along the principal axes of the mass tensor. The method is illustrated by means of some concrete examples. In addition to the illustration of constraint-field correlation functions and how they add up to the mean fields, followed by illustrations of the variance characteristics of field realizations, we concentrate in particular on the consequences of imposing gravitational field constraints (or, equivalent in the linear regime for growing mode fluctuations, peculiar velocity field constraints).

Subject headings: Cosmology : theory – Galaxies: clustering – large-scale structure of the Universe – Methods: numerical

1. Introduction

In the standard scenario of structure formation galaxies and the large-scale structure form through the growth of primordial density perturbations. These perturbations take the form of a homogeneous and isotropic random process. In most cases these cosmological density fields are assumed to be Gaussian random fields.

In these density fields the regions around local maxima and minima are of particular interest during the evolution of the perturbation field. The first collapsed structures form generally near (but are not coincident with, Bertschinger & Jain 1994) density peaks, making density maxima the progenitors of objects like galaxies and clusters. On the other hand, the minima will be the centres of expanding voids. The properties of peaks in Gaussian random fields have been described extensively in the literature. In order to identify an object of a certain size and mass in a Gaussian random field one discards smaller scale objects from consideration. This is achieved by filtering the density field on an appropriate scale to reflect the linear evolution of the proto-objects. While in

some scenarios the filter function is a consequence of a simple phenomenon (e.g. free-streaming of neutrinos in a Hot Dark Matter scenario) in other cases one is forced to invoke an artificial filter to approximate the complicated processes of hierarchical merging (e.g. in the Cold Dark Matter scenario).

A description of the properties of these filtered fields was given by Doroshkevich (1970), Peacock and Heavens (1985), and Bardeen et al. (1986; hereafter BBKS). Beside global parameters such as the number density and spatial correlations of peaks found in these filtered fields they also derived the distribution of their height, shape and orientation. Furthermore, BBKS derived the mean and variance of the density profiles around peaks. As soon as these structures enter the nonlinear regime the coupling of modes breaks down the above approach of filtering. To investigate the further evolution one is therefore forced to resort to N-body simulations. However, in order to follow the evolution of a particular object one needs to be able to start off with a primordial density field containing such an object.

Unfortunately, the methods of BBKS apply only to point processes and cannot be used to construct an actual sample of a density profile around a peak with predetermined parameters such as peak height, shape and orientation. The usual approach is therefore to generate an unconstrained realization of a Gaussian field and then to search for peaks or regions that satisfy the desired constraints. In many instances this is an inefficient approach. For example, giant clusters or voids will be so rare that either many samples have to be generated or that a large box needs to be used to ensure that the object is indeed present in the simulation volume. The latter will yield a severely degraded resolution which conflicts with the desire to describe these objects in as much detail as possible. Similar considerations apply when many properties need to be specified to obtain the desired object, even while the corresponding additional constraints do not represent unlikely values. By being able to specify beforehand some of the properties and to ensure the presence of such a peak or region in the simulation volume the required effort will be minimized. Simultaneously, the resolution will be maximized. Potentially the most important advantage of this approach is that the influence of several physical quantities on the evolution of structures can be studied systematically by generating realizations wherein one or more constraints have various values.

The fundamental theory of these constrained random fields was set forth by Bertschinger (1987). He generalized the treatment used by BBKS to give a full statistical description of a Gaussian random field subjected to constraints. Based on these principles he presented a method to correctly sample the probability distribution of the density field subject to linear constraints. This method, however, is rather elaborate and inefficient in its implementation, involving a simulated annealing technique. Although it is useful for generating initial conditions subject to a few constraints (see e.g. Van de Weygaert & Van Kampen 1993), it quickly becomes prohibitively slow for more than two constraints. Looking for a more efficient procedure, Binney and Quinn (1991) showed that Bertschinger's problem simplifies considerably when the random field is expanded in spherical harmonics rather than in a plane wave basis. In the case of a localised set of constraints, such as the presence and shape of a peak at the centre of the box, the problem can then be solved exactly instead of iteratively. However, their algorithm is essentially restricted to the case of quite localised

constraints defined around an obvious centre of symmetry.

The breakthrough in the construction of constrained random fields came with the publication by Hoffman & Ribak (1991, hereafter HR). They realised that for any constraint that is a linear functional of the field the problem can be solved exactly in an elegant and simple manner, without having to invoke complicated iterative techniques. Their method makes it possible to generate initial conditions for N-body simulations that obey a few hundred constraints, e.g. those imposed by the observable universe (see Ganon & Hoffman 1993).

This paper contains a description of the fundamentals and implementation of a specific cosmological application of the method proposed by Hoffman & Ribak (1991). This application consists of the generation of an initial density and velocity field containing one or more density peaks in a simulation box. Apart from being able to determine the location and the scale of the peak, we can specify the central density of the peak, as well as the compactness, shape and orientation of the density field in the immediate surroundings of the peak. In addition, the total matter distribution can be sculpted such that it subjects the peak to a desired amount of net gravitational and tidal forces. In practice, the computer algorithm generates samples of these constrained Gaussian random fields on a lattice, using Monte Carlo techniques. Nearly all relevant calculations are done in Fourier transform space. Some results of cosmological studies based on these constrained initial conditions are presented by Van Haarlem & Van de Weygaert (1993), Van de Weygaert & Babul (1994, 1995).

In this paper, we start with some basic concepts of Gaussian random fields followed by a treatment of the fundamental theory of constrained Gaussian random fields in section 2. The Hoffman-Ribak method for the construction of constrained random fields is described in section 3, followed by a description of our Fourier space implementation. In section 4, we present our application of this method to the generation of peaks, deriving constraint kernels for the various peak quantities. In addition, we provide prescriptions for the probability of the imposed constraints. A realization of a random density field with a constrained peak is presented in section 5. Specifically, we will focus on the influence of imposing a peculiar acceleration and a tidal field. In section 6, we will conclude with a summary and short discussion.

2. Fundamentals of constrained Gaussian random fields

Although the paper by Hoffman and Ribak presents the essentials of the simple direct method to construct samples of constrained random fields, it does not provide its mathematical background. This can be obtained extending our earlier treatments (Bertschinger 1987, Van de Weygaert 1991). Therefore we will first summarize the necessary mathematical background in the notation employed by HR before we get to the presentation of their method.

2.1 Gaussian random fields: basics

Consider a homogeneous and isotropic random field $f(\mathbf{x})$ with zero mean. The random field is defined by the set of N -point joint probabilities,

$$\mathcal{P}_N = P[f(\mathbf{x}_1), f(\mathbf{x}_2), \dots, f(\mathbf{x}_N)] df(\mathbf{x}_1)df(\mathbf{x}_2) \cdots df(\mathbf{x}_N), \quad (1)$$

that the field f has values in the range $f(\mathbf{x}_j)$ to $f(\mathbf{x}_j) + df(\mathbf{x}_j)$ for each of the $j = 1, \dots, N$, with N an arbitrary integer and specified positions $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N$.

Here we restrict ourselves to the study of Gaussian random fields, whose statistical properties are completely characterised by some power spectrum (spectral density) or its Fourier transform, the autocorrelation function. There are both physical and statistical arguments in favour of the assumption that the primordial density field in the Universe was indeed of this nature. If the very early Universe went through an inflationary phase, quantum fluctuations would generate small-amplitude curvature fluctuations. The resulting density perturbation field is generally a Gaussian random process with a nearly Harrison-Zel'dovich scale-invariant primordial power spectrum. But even while inflation did not occur, the density field $f(\mathbf{x})$ will be nearly Gaussian in the rather general case that its Fourier components $\hat{f}(\mathbf{k})$ are independent and have random phases (cf. Scherrer 1992). The Fourier decomposition of the field at a specific location \mathbf{x} can then be seen as the superposition of a large number of independent random variables that are drawn from the same distribution. By virtue of the central limit theorem the distribution of this field will approach normality, and (at least) for small N the multivariate distribution \mathcal{P}_N is multivariate normal (Gaussian):

$$\mathcal{P}_N = \frac{\exp \left[-\frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N f_i (\mathbf{M}^{-1})_{ij} f_j \right]}{[(2\pi)^N (\det \mathbf{M})]^{1/2}} \prod_{i=1}^N df_i, \quad (2)$$

where \mathbf{M}^{-1} is the inverse of the $N \times N$ covariance matrix \mathbf{M} , the generalisation of the variance σ^2 in a one-dimensional normal distribution. \mathbf{M} is completely determined by the autocorrelation function $\xi(r)$ if the field is a Gaussian random field,

$$M_{ij} \equiv \langle f(\mathbf{x}_i) f(\mathbf{x}_j) \rangle = \xi(\mathbf{x}_i - \mathbf{x}_j) = \xi(|\mathbf{x}_i - \mathbf{x}_j|), \quad (3)$$

Throughout this paper the brackets $\langle \dots \rangle$ denote an ensemble average. The last relation in equation (3) reflects the fact that our field is a homogeneous and isotropic random process. Since we can consider f as an N -dimensional column vector, we can also write the covariance matrix \mathbf{M} in the convenient form

$$\mathbf{M} = \langle f f^t \rangle, \quad (4)$$

with f^t the transpose of f . By taking the limit as $N \rightarrow \infty$ with uniform spatial sampling, the summations appearing in equation (2) may be turned into integrals. The result

$$\mathcal{P}[f] = e^{-S[f]} \mathcal{D}[f], \quad (5)$$

is similar to the quantum-mechanical partition function in path integral form, where S is the action functional. Although there is no direct connection with quantum field theory, S will be referred to as the action. The expression for the action S for a Gaussian random field can be obtained from equation (2),

$$S[f] = \frac{1}{2} \int d\mathbf{x}_1 \int d\mathbf{x}_2 f(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) f(\mathbf{x}_2), \quad (6)$$

where K is the functional inverse of the correlation function ξ ,

$$\int d\mathbf{x} K(\mathbf{x}_1 - \mathbf{x}) \xi(\mathbf{x} - \mathbf{x}_2) = \delta_D(\mathbf{x}_1 - \mathbf{x}_2), \quad (7)$$

and δ_D the Dirac delta function. The measure $\mathcal{D}[f]$ is most easily evaluated on a lattice, where it is just the product of differentials df_i divided by a normalization constant.

Note that we use the notation $\mathcal{P}[f]$ to refer to an infinitesimal probability with measure $\mathcal{D}[f]$; the probability density is $\exp(-S[f])$. The square brackets in $\mathcal{P}[f]$ and $S[f]$ indicate that these are functionals, i.e., they map the complete function $f(\mathbf{x})$ to one number.

To compute expectation values $\langle A \rangle$ of properties (functionals) of the random field $f(\mathbf{x})$, such as the galaxy mass distribution or the distribution of cluster shapes, one integrates the functional over all possible density fields $f(\mathbf{x})$, weighting each by the probability from equation (5),

$$\langle A \rangle = \frac{\int A[f] e^{-S[f]} \mathcal{D}[f]}{\int e^{-S[f]} \mathcal{D}[f]}. \quad (8)$$

This is exactly analogous to the sum over histories or paths in the Feynman path integral formulation of quantum mechanics (Feynman and Hibbs 1965). As in quantum field theory, there are two practical ways to evaluate cosmological path integrals: perturbation series and Monte Carlo integration.

The perturbation series approach to path integrals, based on Feynmann diagrams, is limited to a small number of applications, as it runs into difficulties when cosmological structures become nonlinear. A more general way to evaluate path integrals, which is adopted here, is by Monte Carlo integration. By generating realisations f_i of the density field, and evaluating the corresponding values $A[f_i]$ of the quantity A , the mean of these values is determined:

$$\langle A \rangle = \frac{\sum_i A[f_i]}{\sum_i 1}. \quad (9)$$

The subsequent non-linear evolution is treated by performing N -body simulations of a specific realisation f . The central issue in this method is the need to draw samples f_i which have a probability distribution proportional to $\exp(-S[f])$ (eq. 5).

2.2 Gaussian Random Fields: constraints

The complicating factor in generating Gaussian random density fields subject to one or more constraints is that correlations couple all points of the field with all other points. Therefore, instead of describing the field in terms of an infinite product of one-dimensional probabilities, one is forced to formulate the problem using infinite-dimensional probability spaces (see eq. 5).

The strategy followed by Bertschinger (1987) is to incorporate the set of constraints imposed on the density field $f(\mathbf{x})$ in the action $S[f]$, according to the definition in equation (5). A realization of the constrained density field is then obtained by properly sampling the resulting distribution

function $\exp(-S[f])$. To make clear how the constraints are incorporated in the action, we consider a field $f(\mathbf{x})$ that is subject to a set of M constraints,

$$\Gamma = \{C_i \equiv C_i[f; \mathbf{x}_i] = c_i; \quad i = 1, \dots, M\}. \quad (10)$$

The constraints are therefore imposed by forcing the field $C_i[f; \mathbf{x}]$, ($i = 1, \dots, M$), a functional of the field $f(\mathbf{x})$ as well as a function of the point \mathbf{x} , to have the specific value c_i at the position \mathbf{x}_i . The constraints C_i are assumed to be linear functionals. Examples of such functionals are the value of the field itself at the point \mathbf{x}_α , the derivative of the field $f(\mathbf{x})$ at the point \mathbf{x}_β , or a convolution over $f(\mathbf{x})$ with some function $g(\mathbf{x})$,

$$\begin{aligned} C_\alpha[f; \mathbf{x}_\alpha] &= f(\mathbf{x}_\alpha) = c_\alpha, \\ C_\beta[f; \mathbf{x}_\beta] &= \frac{\partial}{\partial x} f(\mathbf{x})|_{\mathbf{x}_\beta} = c_\beta, \\ C_\gamma[f; \mathbf{x}_\gamma] &= \int g(\mathbf{x}_\gamma - \mathbf{x}) f(\mathbf{x}) d\mathbf{x} = c_\gamma. \end{aligned} \quad (11)$$

The constraints C_α and C_β can be considered as particular cases of a convolution of $f(\mathbf{x})$ with functions g_α and g_β respectively,

$$\begin{aligned} g_\alpha(\mathbf{x}_\alpha - \mathbf{x}) &= \delta_D(\mathbf{x}_\alpha - \mathbf{x}) \\ g_\beta(\mathbf{x}_\beta - \mathbf{x}) &= \frac{\partial}{\partial x} \delta_D(\mathbf{x}_\beta - \mathbf{x}). \end{aligned} \quad (12)$$

A broad class of constraints can be considered as such, so that a treatment of the constraints in the form of a convolution is not a serious restriction. In particular, we will see in section 4 that the expressions for the 10 constraints needed to specify the height, shape and orientation of a peak in the filtered density field $f_F(\mathbf{x})$, the 3 constraints to specify its peculiar acceleration and the 5 constraints to specify its tidal field can all be written as convolutions over the field $f(\mathbf{x})$, with the convolution functions g depending on the kind of constraint.

Since we limit our fields $f(\mathbf{x})$ to those that obey the set of M constraints Γ , the probability of possible realisations $f(\mathbf{x})$ is the conditional probability $\mathcal{P}[f(\mathbf{x})|\Gamma]$,

$$\mathcal{P}[f|\Gamma] = \frac{\mathcal{P}[f, \Gamma]}{\mathcal{P}[\Gamma]} = \frac{\mathcal{P}[f]}{\mathcal{P}[\Gamma]}. \quad (13)$$

The second equality follows because the constraints are linear functionals of f , so that the joint probability space for f and Γ is the same as the probability space for f . Because the constraints C_i are linear functionals the central limit theorem assures them to have a Gaussian probability distribution when applied on a Gaussian field $f(\mathbf{x})$. The covariance matrix \mathbf{Q} of the constraints' probability distribution can be expressed as (cf. eq. 4),

$$\mathbf{Q} = \langle C C^t \rangle, \quad (14)$$

where C is the M -dimensional column vector with elements C_i , and C^t its transpose. The joint probability $\mathcal{P}[\Gamma]$ for the set of constraints Γ is therefore the following multivariate Gaussian distribution (cf. eq. 2),

$$\mathcal{P}[\Gamma] = \frac{\exp\left[-\frac{1}{2} \sum_{i=1}^M \sum_{j=1}^M C_i (\mathbf{Q}^{-1})_{ij} C_j\right]}{[(2\pi)^M (\det \mathbf{Q})]^{1/2}} \prod_{i=1}^M dC_i, \quad (15)$$

or, in a more concise notation,

$$\mathcal{P}[\Gamma] = \exp\left(-\frac{1}{2} C^t \mathbf{Q}^{-1} C\right) \mathcal{D}[\Gamma], \quad (16)$$

where the measure $\mathcal{D}[\Gamma]$ is defined as

$$\mathcal{D}[\Gamma] = \frac{1}{[(2\pi)^M (\det \mathbf{Q})]^{1/2}} \prod_{i=1}^M dC_i. \quad (17)$$

When each field $f(\mathbf{x})$ is represented by its value at N points (e.g. in a discrete computer representation) we can picture the problem in a geometrical way. The fields $f(\mathbf{x})$ can be considered as N -dimensional vectors (f_1, \dots, f_N) . The constraint set Γ carves out an $(N - M)$ -dimensional hypersurface in this N -dimensional vector space, consisting of all fields obeying these constraints. In other words, the set Γ is an $(N - M)$ -dimensional hypersurface, in particular a hyperplane when the constraints are linear.

The expression for the conditional probability of the field $f(\mathbf{x})$ given the set of constraints Γ , $\mathcal{P}[f|\Gamma]$, follows after inserting equations (5), (6) and (16) into equation (13),

$$\mathcal{P}[f|\Gamma] = \exp\left[-\frac{1}{2} \left(\int \int f(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) f(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2 - C^t \mathbf{Q}^{-1} C \right)\right] \frac{\mathcal{D}[f]}{\mathcal{D}[\Gamma]}. \quad (18)$$

This result shows that the constraints Γ are incorporated into the formalism by a change of the action $S[f]$ to

$$2S[f] = \int \int f(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) f(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2 - C^t \mathbf{Q}^{-1} C \quad (19)$$

In Appendix A it is shown that this constrained action may be written in a simple and revealing form,

$$2S[F] = \int d\mathbf{x}_1 \int d\mathbf{x}_2 F(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) F(\mathbf{x}_2), \quad (20)$$

where the residual field $F(\mathbf{x})$ is defined as the difference between a Gaussian field $f(\mathbf{x})$ satisfying the constraint set Γ and the ensemble mean $\bar{f}(\mathbf{x})$ of all these fields,

$$F(\mathbf{x}) \equiv f(\mathbf{x}) - \bar{f}(\mathbf{x}). \quad (21)$$

Figure 1. Illustration of the construction of a constrained random field. The field contains two peaks, an elongated one defined on a Gaussian scale of $4h^{-1}$ Mpc at $[x, y] = [65.0, 65.0] h^{-1}$ Mpc, and a more compact one defined on a Gaussian scale of $2h^{-1}$ Mpc at a position of $[x, y] = [35.0, 35.0] h^{-1}$ Mpc. The corresponding mean constrained field (\bar{f}) is shown in the top left frame, to which the residual field $F = f - \bar{f}$ in the top right frame is added to obtain the constrained random field realization (f) shown in the bottom frames. The left frame shows the field after smoothing with a Gaussian filter with $R_f = 2h^{-1}$ Mpc, while the right frame is the field after smoothing on a scale of $R_f = 4h^{-1}$ Mpc. The fluctuation field has a standard cold dark matter spectrum ($\Omega = 1.0, h = 0.5$).

The conditional probability function can therefore be described as a shifted Gaussian around the ensemble mean field, $\bar{f}(\mathbf{x})$ (see Appendix A),

$$\bar{f}(\mathbf{x}) = \langle f(\mathbf{x}) | \Gamma \rangle = \xi_i(\mathbf{x}) \xi_{ij}^{-1} c_j, \quad (22)$$

where summation over repeated indices is used. Thus, $\bar{f}(\mathbf{x})$ is the “most likely” field satisfying the constraints and it equals the “average density profile” obtained by BBKS. More precisely, $f = \bar{f}$ is a stationary point of the action:

$$\frac{\delta S}{\delta f} = 0 \quad \text{for } f = \bar{f}. \quad (23)$$

In equation (22) $\xi_i(\mathbf{x})$ is the cross-correlation between the field and the i th constraint $C_i[f; \mathbf{x}_i]$ while ξ_{ij} is the $(ij)^{th}$ element of the constraints’ correlation matrix \mathbf{Q} ,

$$\begin{aligned} \xi_i(\mathbf{x}) &= \langle f(\mathbf{x}) C_i \rangle, \\ \xi_{ij} &= \langle C_i C_j \rangle. \end{aligned} \quad (24)$$

If the constraints C_i involve only the field itself at single points, like C_α in equation (11), both the correlation matrix ξ_{ij} and $\xi_i(\mathbf{x})$ reduce to the two-point correlation function $\xi(\mathbf{x})$,

$$\begin{aligned} \xi_i(\mathbf{x}) &= \langle f(\mathbf{x}) f(\mathbf{x}_i) \rangle = \xi(|\mathbf{x}_i - \mathbf{x}|), \\ \xi_{ij} &= \langle f(\mathbf{x}_i) f(\mathbf{x}_j) \rangle = \xi(|\mathbf{x}_i - \mathbf{x}_j|). \end{aligned} \quad (25)$$

In effect, the residual field $F(\mathbf{x})$ provides random noise which is added to the signal $\bar{f}(\mathbf{x})$, which is completely fixed by the imposed set of constraints Γ . Generating a sample $f(\mathbf{x})$ obeying the constraints $\{C_i[f; \mathbf{x}_i] = c_i; \quad i = 1, \dots, M\}$ therefore consists of constructing \bar{f} from C_i and c_i according to equation (22), subsequently generating the noise $F(\mathbf{x})$, and adding them:

$$f(\mathbf{x}) = \bar{f}(\mathbf{x}) + F(\mathbf{x}) = \xi_i(\mathbf{x}) \xi_{ij}^{-1} c_j + F(\mathbf{x}). \quad (26)$$

Notice that the residual field F is a Gaussian field because it is the difference between two Gaussian fields. The whole problem of constructing a constrained random field has now been reduced to a proper sampling of F . This is complicated by the fact that $F(\mathbf{x})$ is not entirely random but subject to the set of M constraints Γ_0 :

$$\Gamma_0 \equiv \{C_i[f; \mathbf{x}_i] = 0; \quad i = 1, \dots, M\}. \quad (27)$$

This follows directly from the fact that the constraints C_i are linear functionals and F is the difference between two fields,

Figure 2. Linear density profiles along the central x -axis of the field shown in figure 1. On the left, the field has been smoothed using a Gaussian filter with $R_f = 2h^{-1}$ Mpc. The solid line shows the constrained field (f). The dotted line is the mean field (\bar{f}) and the dashed line the residual field $F = f - \bar{f}$. On the right the same field, but now after filtering on a scale of $4h^{-1}$ Mpc.

An illustration of the sketched constrained field construction procedure, based on equation (26), is provided by figure 1. Note that both the original Bertschinger prescription (1987) and the Hoffman-Ribak procedure (1991) are based on this equation (the particular realization in figure 1 has been generated with the Hoffman-Ribak code described in this paper). The fluctuation field in the $100h^{-1}$ Mpc box has a standard cold dark matter spectrum ($\Omega = 1.0, h = 0.5$) and contains two peaks of different shape and scale, a spherical $4\sigma_0(2h^{-1}$ Mpc) overdensity and an elongated $3\sigma_0(4h^{-1}$ Mpc) overdensity. Density contour maps (filtered on a scale of $2h^{-1}$ Mpc) of the mean field \bar{f} defined by this constraint (top left), an accompanying residual field realisation F (top right) and the resulting constrained field f (bottom left) are shown in slices of width $1/20th$ of the boxsize taken along the z -direction. The slices pass through the centre of the box. Figure 1d shows the constrained field f smoothed on a scale of $4h^{-1}$ Mpc. A good idea of the relative amplitudes of the mean, residual and constrained field in figure 1 can be obtained from linear density profiles through the density field. Figure 2 shows such profiles, taken along the central x -axis, passing through the outskirts of both peaks. The left figure corresponds to the density field at a Gaussian filtering scale of $2h^{-1}$ Mpc, while the right figure has a Gaussian smoothing scale of $4h^{-1}$ Mpc. The dotted line is the mean field \bar{f} , the dashed line the residual field F , and the solid line the superposition of the two, the constrained field realization f .

$$C_i[F] = C_i[f - \bar{f}] = C_i[f] - C_i[\bar{f}] = c_i - c_i = 0. \quad (28)$$

This fact is independent of the numerical values $\{c_i\}$ of the constraints Γ imposed on the field $f(\mathbf{x})$.

3. Sampling constrained Gaussian random fields

Application of the construction procedure based on equation (26) requires the ability to properly sample $\exp(-S[F])$ for the random field $F(\mathbf{x})$. The sampling procedure forms the core of any constrained random field algorithm, and determines its effectiveness and reliability. The sampling is carried out most conveniently in Fourier space, where the action $S[F]$ is diagonalized (appendix B),

$$S[F] = \int \frac{d\mathbf{k}}{(2\pi)^3} \frac{|\hat{F}(\mathbf{k})|^2}{2P(k)}, \quad (29)$$

where $\hat{F}(\mathbf{k})$ is the Fourier transform of the residual field $F(\mathbf{x})$,

$$F(\mathbf{x}) = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{F}(\mathbf{k}) e^{-i\mathbf{k}\cdot\mathbf{x}}, \quad (30)$$

and $P(k)$ the power spectrum of the field (see eq. 41 for the formal definition). Note that in this paper we adopt a different Fourier transform than Bertschinger (1987, 1992).

In the case of an unconstrained field, for which $F(\mathbf{x}) = f(\mathbf{x})$, all harmonics $\hat{F}(\mathbf{k})$ are mutually independent and normally distributed. This makes sampling relatively easy. However, for a constrained field the residual field is subject to the constraints Γ_0 (eq. 27), so that its Fourier components are no longer mutually independent. The coupling of the different Fourier modes turns the sampling of the action $S[F]$ into a non-trivial problem. In earlier work we (Bertschinger 1987, Van de Weygaert 1991) accomplished the sampling of the residual field, carried out in discrete Fourier space $\hat{F}(\mathbf{k}_j)$, by means of an iterative “simulated annealing” technique. The action was sampled by means of a Markov chain, starting with an initial guess for the harmonics and updating them iteratively, each update depending only on values of the most recent estimate. After a number of iterations the Markov chain relaxes to a steady state with \hat{F} correctly sampling the action. The algorithm used for the update is the “heat bath” algorithm, which treats the discrete set of harmonics \hat{F} like a series of coupled oscillators in thermal contact with a heat bath of fixed temperature. The

heat bath generates random fluctuations in each harmonic which couple to all other harmonics. The fluctuations drive the system towards a state of “thermal” equilibrium in which the action is distributed properly. The algorithm requires $\mathcal{O}[(M^2 + 1)N]$ operations to generate one independent realisation, where N is the number of degrees of freedom (roughly the number of grid points for the density) and M is the number of constraints. A disadvantage of this iterative approach is that as the grid density grows and as the number of constraints increases to more than a few, the system “anneals” so slowly that the algorithm becomes prohibitively expensive and impractical. Additionally, there is no unique way of deciding at which stage the system has annealed to the desired equilibrium.

3.1 Hoffman-Ribak Algorithm

The crucial observation by Hoffman & Ribak (1991) is that the residual field $F(\mathbf{x})$ has some unique properties which simplify the construction of a realisation of a constrained field substantially. While it was already known that the mean value of $F(\mathbf{x})$ is independent of the numerical values c_i of the constraints Γ ,

$$\langle F(\mathbf{x}) | \Gamma \rangle = \langle f(\mathbf{x}) - \bar{f}(\mathbf{x}) | \Gamma \rangle = \langle f(\mathbf{x}) | \Gamma \rangle - \bar{f}(\mathbf{x}) = 0, \quad (31)$$

it had not been realized earlier that this is true for the complete probability distribution $\mathcal{P}[F|\Gamma]$ of the residual field $F(\mathbf{x})$ itself (see appendix C), i.e.

$$\mathcal{P}[F|\Gamma_1] = \mathcal{P}[F|\Gamma_2] \quad \text{for all } \Gamma_1, \Gamma_2. \quad (32)$$

The observation that the statistical properties of the residual field $F(\mathbf{x})$ are all *independent* of the numerical values c_i is the key element of the Hoffman-Ribak method, rendering unnecessary a direct sampling from the complicated action $S[F]$. A particular residual field $F(\mathbf{x})$ can as well have been sampled from the set of fields subject to the constraints Γ as from the fields belonging to some arbitrary constraint set $\tilde{\Gamma}$. The residual field $\tilde{F}(\mathbf{x})$ that is obtained by generating an unconstrained realisation $\tilde{f}(\mathbf{x})$ of the field, and subtracting the mean field $\bar{\tilde{f}}$ of the constraint set $\tilde{\Gamma}$ to which it belongs, is therefore a correctly sampled residual field for the constraint set Γ .

These considerations lead to the following strategy for constructing a constrained realisation of the field $f(\mathbf{x})$, consisting of five stages:

- (1) Create a random, unconstrained, realisation $\tilde{f}(\mathbf{x})$, a homogeneous and isotropic Gaussian random field whose statistics are determined by the power spectrum alone.
- (2) Calculate for this particular realisation $\tilde{f}(\mathbf{x})$ the values \tilde{c}_i of the constraints $\{C_i(\mathbf{x})|_{\mathbf{x}_i}, i = 1, \dots, M\}$. These variables define a set of constraints, $\tilde{\Gamma} = \{\tilde{c}_i\}$.
- (3) Calculate for this “random” constraint set $\tilde{\Gamma}$ the corresponding mean field, using

$$\bar{\tilde{f}}(\mathbf{x}) = \langle \tilde{f}(\mathbf{x}) | \tilde{\Gamma} \rangle = \xi_i(\mathbf{x}) \xi_{ij}^{-1} \tilde{c}_j. \quad (33)$$

- (4) Evaluate the residual field \tilde{F} of the random realisation:

$$\tilde{F}(\mathbf{x}) = \tilde{f}(\mathbf{x}) - \bar{\tilde{f}}(\mathbf{x}). \quad (34)$$

This residual field \tilde{F} can also be considered the residual field of a particular realisation subject to the desired constraints, Γ .

- (5) Evaluate the desired mean field $\bar{f}(\mathbf{x})$, using equation (22), and add it to the residual field $\tilde{F}(\mathbf{x})$ (eq. 34) to obtain a particular realisation of the desired constrained Gaussian random field $f(\mathbf{x})$:

$$f(\mathbf{x}) = \tilde{f}(\mathbf{x}) + \xi_i(\mathbf{x}) \xi_{ij}^{-1} (c_j - \bar{c}_j) \quad (35)$$

The field $f(\mathbf{x})$ constructed in this way obeys the constraints and replaces the unconstrained field $\tilde{f}(\mathbf{x})$. Note that there is a one-to-one correspondence between the trial field $\tilde{f}(\mathbf{x})$ and $f(\mathbf{x})$. Furthermore, the ensemble of realisations produced by the algorithm presented here properly samples the subensemble of all realisations constrained by Γ . The algorithm is optimal because it is exact and involves only one realisation of an unconstrained random field and the calculation of the mean field under the given constraints.

3.2 The practical implementation

Our implementation of the Hoffman-Ribak algorithm has two important elements. Firstly, for reasons of convenience, all necessary calculations are carried out in Fourier space. Secondly, the constrained field $f(\mathbf{x})$ is generated on a periodic three-dimensional lattice of side L , so that $f(\mathbf{x})$ is evaluated on $N(\propto L^3)$ gridpoints. The result can be considered to be an $N(\propto L^3)$ vector $f = [f(\mathbf{x}_1), \dots, f(\mathbf{x}_N)]$.

The central equation of the Hoffman-Ribak algorithm for generating a constrained field realization $f(\mathbf{x})$ is equation (35). We assume that, as in the case of the 18 peak constraints (section 4), the M constraints $C_i[f; \mathbf{x}_i] = c_i$ on the field $f(\mathbf{x})$ are convolutions of the field $f(\mathbf{x})$ with some kernel $H_i(\mathbf{x}; \mathbf{x}_i)$,

$$C_i[f; \mathbf{x}_i] = \int d\mathbf{x} H_i(\mathbf{x}; \mathbf{x}_i) f(\mathbf{x}) = c_i. \quad (36)$$

In the case of the peak constraints on the local density field (section 4.2) the convolution kernel is a Gaussian filter function or one of its first or second derivatives.

The Fourier transforms of the field $f(\mathbf{x})$ and the kernel $H_i(\mathbf{x}; \mathbf{x}_i)$ are defined by

$$\begin{aligned} f(\mathbf{x}) &= \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{f}(\mathbf{k}) e^{-i\mathbf{k}\cdot\mathbf{x}}, \\ H_i(\mathbf{x}; \mathbf{x}_i) &= \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{H}_i(\mathbf{k}) e^{-i\mathbf{k}\cdot\mathbf{x}}. \end{aligned} \quad (37)$$

Consequently, Parseval's theorem yields the following Fourier expression for the constraint $C_i[f; \mathbf{x}_i] = c_i$,

$$C_i[f; \mathbf{x}_i] = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{H}_i^*(\mathbf{k}) \hat{f}(\mathbf{k}) = c_i, \quad (38)$$

The constraint's correlation function ξ_{ij} can be evaluated by using equation (38),

$$\begin{aligned}\xi_{ij} &\equiv \langle C_i[f; \mathbf{x}_i] C_j[f; \mathbf{x}_j] \rangle = \left\langle \int \frac{d\mathbf{k}_1}{(2\pi)^3} \hat{H}_i^*(\mathbf{k}_1) \hat{f}(\mathbf{k}_1) \int \frac{d\mathbf{k}_2}{(2\pi)^3} \hat{H}_j(\mathbf{k}_2) \hat{f}^*(\mathbf{k}_2) \right\rangle \\ &= \int \frac{d\mathbf{k}_1}{(2\pi)^3} \frac{d\mathbf{k}_2}{(2\pi)^3} \hat{H}_i^*(\mathbf{k}_1) \hat{H}_j(\mathbf{k}_2) \langle \hat{f}(\mathbf{k}_1) \hat{f}^*(\mathbf{k}_2) \rangle.\end{aligned}\quad (39)$$

This immediately leads to the Fourier integral expression

$$\xi_{ij} = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{H}_i^*(\mathbf{k}) \hat{H}_j(\mathbf{k}) P(k), \quad (40)$$

where we have used Bertschinger's definition (1992) for the spectral density $P(k)$, modified by a factor $(2\pi)^3$ owing to our different Fourier transform convention,

$$(2\pi)^3 P(k_1) \delta_D(\mathbf{k}_1 - \mathbf{k}_2) = \langle \hat{f}(\mathbf{k}_1) \hat{f}^*(\mathbf{k}_2) \rangle, \quad (41)$$

with $\delta_D(\mathbf{k}_1 - \mathbf{k}_2)$ the Dirac delta function. Once the expression for $P(k)$ and the Fourier transform $\hat{H}_i(\mathbf{k})$ of the constraint kernel are known, ξ_{ij} can be easily calculated from equation (40).

In a similar way we obtain an expression for the cross-correlation between the field and the i^{th} constraint, $\xi_i(\mathbf{x})$,

$$\begin{aligned}\xi_i(\mathbf{x}) &\equiv \langle f(\mathbf{x}) C_i[f; \mathbf{x}_i] \rangle = \left\langle \int \frac{d\mathbf{k}_1}{(2\pi)^3} \hat{f}(\mathbf{k}_1) e^{-i\mathbf{k}_1 \cdot \mathbf{x}} \int \frac{d\mathbf{k}_2}{(2\pi)^3} \hat{H}_i(\mathbf{k}_2) \hat{f}^*(\mathbf{k}_2) \right\rangle \\ &= \int \frac{d\mathbf{k}_1}{(2\pi)^3} \frac{d\mathbf{k}_2}{(2\pi)^3} \langle \hat{f}(\mathbf{k}_1) \hat{f}^*(\mathbf{k}_2) \rangle \hat{H}_i(\mathbf{k}_2) e^{-i\mathbf{k}_1 \cdot \mathbf{x}},\end{aligned}\quad (42)$$

which in combination with the definition of the spectral density (eq. 41) yields the expression

$$\xi_i(\mathbf{x}) = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{H}_i(\mathbf{k}) P(k) e^{-i\mathbf{k} \cdot \mathbf{x}}. \quad (43)$$

Inserting this expression into equation (35) leads to the following Fourier integral expression for the constrained field,

$$\begin{aligned}f(\mathbf{x}) &= \tilde{f}(\mathbf{x}) + \xi_i(\mathbf{x}) \xi_{ij}^{-1} (c_j - \tilde{c}_j) \\ &= \int \frac{d\mathbf{k}}{(2\pi)^3} \left[\hat{\tilde{F}}(\mathbf{k}) + P(k) \hat{H}_i(\mathbf{k}) \xi_{ij}^{-1} (c_j - \tilde{c}_j) \right] e^{-i\mathbf{k} \cdot \mathbf{x}}.\end{aligned}\quad (44)$$

The only element left in the calculation of the constrained realization $f(\mathbf{x})$ is the unconstrained field $\tilde{f}(\mathbf{x})$. As was noted above, $\tilde{f}(\mathbf{x})$ is most conveniently generated in Fourier space, where its Fourier components $\hat{\tilde{F}}(\mathbf{k})$ are mutually independent and Gaussian distributed.

In practice the above expressions are evaluated on a three dimension grid of $N(\propto L^3)$ gridpoints, and the corresponding Fourier integrals are replaced by discrete Fourier sums. Summarizing, the process of setting up a constrained field for a given spectrum $P(k)$ consists of four steps. Firstly, the value of the constraint kernel is evaluated on N Fourier gridpoints \mathbf{k}_i , an $\mathcal{O}(N)$ operation. Secondly, the matrix ξ_{ij} is calculated by means of equation (40), with a total computational cost

proportional to $\mathcal{O}(M^2N)$, after which its inverse is determined, a $\mathcal{O}(M^3)$ procedure. Subsequently, the N unconstrained field components $\hat{\hat{F}}(\mathbf{k})$ are generated, from which the value of the corresponding constraint values \tilde{c}_i are evaluated using equation (38). The computational cost of the latter is $\mathcal{O}(MN)$. Finally, the constrained field f is determined from equation (44), consisting of the $\mathcal{O}(M^2N)$ evaluation of the products $\hat{H}_i(\mathbf{k})\xi_{ij}^{-1}(c_j - \tilde{c}_j)$ for all wavenumbers \mathbf{k} , followed by a Fourier transform of cost $\mathcal{O}(N \log N)$. Thus, the total cost is $\mathcal{O}[(M^2 + \log N)N]$ (the cost of inverting the constraints is negligible because $N \gg M$). Although this scaling is no better than the $\mathcal{O}[(M^2+1)N]$ scaling of the iterative heat bath method (Bertschinger 1987), the coefficient of proportionality is much smaller because no iteration is required.

4. The peak constraints

An important cosmological application for a constrained random field algorithm is the generation of an initial density field containing one or more peaks (or, equivalently, dips). A peak is identified as a local maximum in the density field that has been smoothed by some filter function or, more generally, as the immediate surroundings of this maximum. The choice of the filter will depend on the specific application. The scale of the peak is defined to be the characteristic scale of that filter function. Depending on their scale, these density peaks may be the progenitors of galaxies, clusters or superclusters. The constrained random field algorithm makes it possible to specify the height, compactness, shape and orientation of the density field in the immediate vicinity of the peak, while the total matter distribution can be sculpted such that the peak is subjected to a desired amount of net gravitational and tidal forces. In the linear clustering regime these forces are directly proportional to the peculiar velocity of the peak and the components of the shear at its location.

Unlike the other constraints, the four quantities to describe the position and scale of the peak are not imposed via the algorithm described in the previous section. Rather, they are parameters that enter via the kernels $H_i(\mathbf{x}; \mathbf{x}_i)$ (eq. 36) of each of the constraints. In addition to its scale and location, a peak in the smooth density field is specified by 18 constraints. The height of the peak needs to be specified while 3 constraints are needed to ensure that the 3 first derivatives of the smooth density field vanish at its summit. The 6 second-order derivatives of the density field are set by specifying the compactness, the axis ratios and the orientation of the peak. These 10 constraints together determine the density distribution in the immediate vicinity of the peak. The specification of the gravitational field around the peak introduces 8 additional constraints: The 3 components of the smoothed peculiar acceleration at the location of the peak and the 5 independent components of the traceless tidal field tensor.

The constraints C_i that we use in our peak algorithm are a combination of one or more of the above quantities. Once a constraint has been specified an expression for the corresponding kernel H_i is derived (see eq. 36). In the practical implementation we derive the expression for the Fourier transform of H_i , $\hat{H}_i(\mathbf{k})$. By working directly in Fourier space we save one FFT and at the same time guarantee a higher accuracy of the results.

After an initial phase of linear evolution in which the Zel'dovich (1970) approximation is used, the further non-linear evolution of the matter distribution surrounding the peak is usually followed

by N-body simulations. It is evident that the use of the constrained random field code makes it possible to study the formation and evolution of these objects more systematically than possible with the conventional methods based on unconstrained fields. Among others, this will provide considerably more insight into the question of which physical parameters and processes have the largest influence on the fate of an object.

In the following we will drop the explicit time dependence in our notation. The value of each of the quantities will be the value that the quantity has when it is linearly extrapolated towards the expansion factor a (with $a = 1$ the present epoch). The treatment in the next sections will be in comoving coordinates and wavevectors, while all spatial derivatives are with respect to these comoving coordinates.

4.1 Peak scale and position

Many cosmological studies have assumed that present-day nonlinear object like galaxies or clusters are the result of the collapse of peaks in the primordial density fields whose height exceeds some threshold, after having smoothed the field with a filter of a certain shape and scale. Because many cosmological scenarios do not possess a natural filtering scale, often an ad hoc filter has to be invoked to define the objects. In this paper we use a Gaussian filter because of its simplicity and smoothing properties. However, the formalism is equally valid for any other filter, and it is trivial to modify the equations (or our computer program) correspondingly.

Although the precise relation between the Gaussian filtering scale R_G and the characteristic mass M_{pk} of a particular object in the present universe is unclear — indeed, the one-to-one association between objects and density peaks is questioned by recent works (Katz, Quinn & Gelb 1993, Bertschinger & Jain 1994, Van de Weygaert & Babul 1994) — we can estimate a reasonable choice using a simple argument. The total mass enclosed by a Gaussian smoothing function with filtering scale R_G in a homogeneous Einstein-de Sitter universe of density $\bar{\rho}$ is

$$M_{pk}(R_G) = (2\pi)^{3/2} \bar{\rho} R_G^3 = 4.3718 \times 10^{12} R_G^3 h^{-1} M_\odot, \quad (45)$$

where R_G is in units of h^{-1} Mpc. For example, if we take for M_{pk} the typical mass of the core of a cluster, $M_c = 6 \times 10^{14} M_\odot$, this yields a Gaussian filter scale of $R_G \approx 4 h^{-1}$ Mpc. Similarly, a radius of $R_G \approx 0.6 h^{-1}$ Mpc corresponds to a mass of $\approx 10^{12} M_\odot$, comparable to the mass of a galaxy with a luminosity equal to L_* if $\Omega = 1$.

The use of the filter function W_G serves a twofold purpose in our peak constraint algorithm. In addition to defining the scale of the peaks in the density field $\rho(\mathbf{x})$ it is also of vital importance in the derivation of the kernels H_i of each of the constraints. The expressions for these kernels are found by using the fact that the peaks are maxima in the filtered density field $f_G(\mathbf{x})$,

$$f_G(\mathbf{x}) = \int d\mathbf{y} f(\mathbf{y}) W_G(\mathbf{y}, \mathbf{x}), \quad (46)$$

where $f(\mathbf{x})$ is the density contrast field,

$$f(\mathbf{x}) = \frac{\rho(\mathbf{x}) - \bar{\rho}}{\bar{\rho}}, \quad (47)$$

(in this equation $\bar{\rho}$ is the average density of the Universe). This convolution integral is equivalent to the Fourier integral

$$\begin{aligned} f_G(\mathbf{x}) &= \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{f}(\mathbf{k}) \hat{W}^*(\mathbf{k}; \mathbf{x}) \\ &= \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{f}(\mathbf{k}) \hat{W}^*(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}} \end{aligned} \quad (48)$$

where $\hat{W}(\mathbf{k}; \mathbf{x})$ and $\hat{W}(\mathbf{k})$ are the Fourier transforms of $W_G(\mathbf{y}, \mathbf{x})$ and $W_G(\mathbf{x}, \mathbf{0})$. In the case of a Gaussian filter,

$$W_G(\mathbf{y}, \mathbf{x}) = \frac{1}{(2\pi R_G^2)^{3/2}} \exp\left(-\frac{|\mathbf{y} - \mathbf{x}|^2}{2R_G^2}\right), \quad (49)$$

$\hat{W}(\mathbf{k}; \mathbf{x})$ and $\hat{W}(\mathbf{k})$ are

$$\hat{W}(\mathbf{k}) = e^{-k^2 R_G^2/2} \quad \text{and} \quad \hat{W}(\mathbf{k}, \mathbf{x}) = \hat{W}(\mathbf{k}) e^{i\mathbf{k} \cdot \mathbf{x}} = e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}}. \quad (50)$$

Note that the position \mathbf{x} of an object causes a phase shift $\mathbf{k} \cdot \mathbf{x}$ with respect to an object that is situated at the origin, $\mathbf{0}$.

4.2 The local density field

Locally, the density field around a peak at position \mathbf{x}_d can be described by the second order Taylor expansion of the density profile $f_G(\mathbf{x})$ around the peak,

$$f_G(\mathbf{x}) = f_G(\mathbf{x}_d) + \frac{1}{2} \sum_{i,j=1}^3 \frac{\partial^2 f_G}{\partial x_i \partial x_j}(\mathbf{x}_d) (x_i - x_{d,i})(x_j - x_{d,j}). \quad (51)$$

In this expansion we have used the fact that the three first derivatives of the field $f_G(\mathbf{x})$ at the location of the local maximum, \mathbf{x}_d , are equal to zero. The equation shows that the requirement that the smoothed density field $f_G(\mathbf{x})$ has a maximum of a certain height, shape, orientation at location \mathbf{x}_d translates into constraints on the value of the smooth density field f_G at \mathbf{x}_d , on its gradient ∇f_G and on the second derivative tensor of the field, $\nabla_i \nabla_j f_G$. This implies that 10 constraints are required to fully specify the local density field around a peak. Also note that the quadratic part of equation (51) should be negative definite if $f_G(\mathbf{x}_d)$ is a maximum. Consequently, the isodensity surfaces $f_G = F$ around the peak are triaxial ellipsoids, whose orientation and size depends on the value of the second derivatives of f_G .

The first constraint is the height of the peak, $f_G(\mathbf{x}_d)$. Usually it is expressed in units of the variance $\sigma_0(R_G) = \langle f_G f_G \rangle^{1/2}$ of the smoothed density field,

$$f_G(\mathbf{x}_d) = \nu_c \sigma_0(R_G), \quad (52)$$

which in combination with the convolution expression for f_G in equation (46) yields the following expression,

$$\int \frac{d\mathbf{k}}{(2\pi)^3} \hat{f}(\mathbf{k}) \hat{W}^*(\mathbf{k}) e^{-i\mathbf{k}\cdot\mathbf{x}_d} = \nu_c \sigma_0(R_G). \quad (53)$$

Consequently, the corresponding constraint kernel $\hat{H}_1(\mathbf{k})$ (see eq. 38) and constraint value c_1 are given by

$$\hat{H}_1(\mathbf{k}) = \hat{W}(\mathbf{k}) e^{i\mathbf{k}\cdot\mathbf{x}_d}, \quad c_1 = \nu_c \sigma_0(R_G). \quad (54)$$

For reasons of clarity and convenience a compilation of the kernels of all peak constraints is given in appendix F.

Three additional constraints are obtained from the extremum demand that the first order derivatives of f_G should be 0 at the peak position \mathbf{x}_d ,

$$\frac{\partial f_G}{\partial x_i}(\mathbf{x}_d) = 0, \quad i = 1, 2, 3. \quad (55)$$

The Fourier expression for the gradient $\nabla f_G(\mathbf{x}_d)$ is obtained by partial differentiation of the integrand in the convolution equation (48),

$$\frac{\partial f_G}{\partial x_j} = \frac{\partial}{\partial x_j} \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{f}(\mathbf{k}) \hat{W}^*(\mathbf{k}) e^{-i\mathbf{k}\cdot\mathbf{x}_d} = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{f}(\mathbf{k}) \hat{W}^*(\mathbf{k}) \frac{\partial}{\partial x_j} (e^{-i\mathbf{k}\cdot\mathbf{x}_d}). \quad (56)$$

This yields the following constraint expressions,

$$\int \frac{d\mathbf{k}}{(2\pi)^3} -i\mathbf{k} \hat{f}(\mathbf{k}) \hat{W}^*(\mathbf{k}) e^{-i\mathbf{k}\cdot\mathbf{x}_d} = 0. \quad (57)$$

The corresponding kernels $\hat{H}_2(\mathbf{k})$, $\hat{H}_3(\mathbf{k})$ and $\hat{H}_4(\mathbf{k})$, and the constraint values c_2 , c_3 and c_4 are therefore

$$\hat{H}_j(\mathbf{k}) = ik_l \hat{W}(\mathbf{k}) e^{i\mathbf{k}\cdot\mathbf{x}_d}, \quad c_j = 0, \quad (58)$$

where $j = 2, \dots, 4$ and the corresponding $l = j - 1$ (also see appendix F).

Finally, there are six constraints that correspond to the shape, compactness, and orientation of the density field around the peak. Because the density field in its vicinity is ellipsoidal (see appendix E), its shape is fully characterized by the two axis ratios $a_{12} \equiv a_1/a_2$ and $a_{13} \equiv a_1/a_3$. The quantity that describes the compactness, or steepness, of the density profile around a peak is the Laplacian $\nabla^2 f_G(\mathbf{x}_d)$. Usually this Laplacian is expressed in units of $\sigma_2(R_G) = \langle \nabla^2 f_G \nabla^2 f_G \rangle^{1/2}$ (see appendix E),

$$\nabla^2 f_G(\mathbf{x}_d) = -x_d \sigma_2(R_G). \quad (59)$$

The minus sign in this definition of x_d is introduced in order for x_d to be negative in the case of a dip and positive for a peak. The orientation of the peak with respect to the coordinate axes is most conveniently specified by the three Euler angles α , β and ψ . The corresponding transformation matrix A_{ij} is given by,

$$A = \begin{pmatrix} \cos \alpha \cos \psi - \cos \beta \sin \alpha \sin \psi & \sin \alpha \cos \psi + \cos \beta \cos \alpha \sin \psi & \sin \beta \sin \psi \\ -\cos \alpha \sin \psi - \cos \beta \sin \alpha \cos \psi & -\sin \alpha \sin \psi + \cos \beta \cos \alpha \cos \psi & -\sin \beta \cos \psi \\ \sin \beta \sin \alpha & -\sin \beta \cos \alpha & \cos \beta \end{pmatrix}. \quad (60)$$

The above six quantities (a_{12} , a_{13} , x_d , α , β and ψ) constrain the six second order derivatives of f_G via the combination (see appendix E for a derivation),

$$\frac{\partial^2 f_G}{\partial x_i \partial x_j} = - \sum_{k=1}^3 \lambda_k A_{ki} A_{kj}, \quad i, j = 1, 2, 3, \quad (61)$$

where A_{ij} are the elements of the orientation matrix (eq. 60), and the λ_i are the eigenvalues of the matrix $-\nabla_i \nabla_j f_G$. The values of λ_i are obtained from the axis ratios a_{12} and a_{13} of the isodensity ellipsoids around the peak, as well as from the steepness of the density profile, x_d , via the relations

$$\lambda_1 = \frac{x_d \sigma_2(R_G)}{(1 + a_{12}^2 + a_{13}^2)}, \quad \lambda_2 = \lambda_1 a_{12}^2, \quad \lambda_3 = \lambda_1 a_{13}^2, \quad (62)$$

A Fourier expression for the second order derivatives of $f_G(\mathbf{x})$ is obtained by double partial differentiation of the integrand of the convolution integral (eq. 48),

$$\int \frac{d\mathbf{k}}{(2\pi)^3} -k_i k_j \hat{f}(\mathbf{k}) \hat{W}^*(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}_d} = - \sum_{k=1}^3 \lambda_k A_{ki} A_{kj},$$

so that we find

$$\hat{H}_l(\mathbf{k}) = -k_i k_j \hat{W}(\mathbf{k}) e^{i\mathbf{k} \cdot \mathbf{x}_d}, \quad c_l = - \sum_{k=1}^3 \lambda_k A_{ki} A_{kj}, \quad (63)$$

for the kernels $\hat{H}_l(\mathbf{k})$ and constraint values c_l , with $l = 5, \dots, 10$ and $i, j = 1, \dots, 3$ (see appendix F for the correct numbering).

4.3 The local gravitational field

The peak constraints that were introduced and discussed in section 4.2 describe the density field in the immediate surroundings of the peak. Of more fundamental importance to the dynamics of a region of space are constraints on the gravitational potential perturbations. The local potential perturbation $\phi(\mathbf{x})$ is the weighted sum of all density perturbations throughout the universe. Constraints on the local potential therefore have immediate repercussions for the global matter distribution. Since we wish to neglect the potential fluctuations on scales smaller than the objects we are interested in, we consider the smoothed potential perturbation field ϕ_G ,

$$\phi_G(\mathbf{x}) = \int d\mathbf{y} \phi(\mathbf{y}) W_G(\mathbf{y}, \mathbf{x}). \quad (64)$$

In our implementation we use a Gaussian function for the filter $W_G(\mathbf{y}, \mathbf{x})$, as in the case of the density field.

It is physically appealing to impose constraints on the potential ϕ via constraints on its derivatives, in particular the gravitational acceleration and the tidal field. The peculiar gravitational acceleration \mathbf{g} at the position \mathbf{x} is

$$\mathbf{g}(\mathbf{x}, t) = \frac{1}{a} \frac{d(a\mathbf{v})}{dt} = -\frac{1}{a} \nabla \phi, \quad (65)$$

where a is the cosmological expansion factor and $\mathbf{v}(\mathbf{x}, t)$ the peculiar velocity of the patch of matter at physical position $\mathbf{r}(t) = a\mathbf{x}(t)$,

$$\mathbf{v} = \frac{d\mathbf{r}}{dt} - H\mathbf{r}. \quad (66)$$

A first-order Taylor expansion of the gravitational field around the peak shows that the dynamical state of the patch of matter in its immediate neighbourhood, on scales larger than the filter scale R_G , is completely specified by the bulk acceleration $\mathbf{g}_G(\mathbf{x}_d) = -\nabla \phi_G/a$, the divergence $\nabla \cdot \mathbf{g}_G$ and by the traceless (comoving) tidal tensor $E_{G,ij}$,

$$g_{G,i}(\mathbf{x}) = g_{G,i}(\mathbf{x}_d) + a \sum_{j=1}^3 \left\{ \frac{1}{3a} (\nabla \cdot \mathbf{g}_G)(\mathbf{x}_d) \delta_{ij} - E_{ij} \right\} (x_j - x_{d,j}), \quad (67)$$

where δ_{ij} is the Kronecker delta and $E_{G,ij}$ is the trace-free part of $-\partial g_{G,i}/\partial r_j = \partial^2 \phi_G/\partial r_i \partial r_j$ (note that here we choose to use physical coordinates r_i , since we are dealing with physical quantities),

$$E_{G,ij} \equiv -\frac{1}{2a} \left\{ \frac{\partial g_{G,i}}{\partial x_i} + \frac{\partial g_{G,j}}{\partial x_j} \right\} + \frac{1}{3a} (\nabla \cdot \mathbf{g}_G) \delta_{ij} = \frac{1}{a^2} \left\{ \frac{\partial^2 \phi_G}{\partial x_i \partial x_j} - \frac{1}{3} \nabla^2 \phi_G \delta_{ij} \right\}. \quad (68)$$

The divergence $\nabla \cdot \mathbf{g}_G/a$ is the component of the gravitational field corresponding to pure radial infall into (or outflow from) the peak. Through the Poisson equation this quantity is directly proportional to the local density perturbation $f_G(\mathbf{x})$,

$$\frac{\nabla \cdot \mathbf{g}_G}{a} = -\frac{1}{a^2} \nabla^2 \phi_G = -\frac{3}{2} \Omega H^2 f_G(\mathbf{x}). \quad (69)$$

The expression for the constraint on $\nabla \cdot \mathbf{g}_G/a$ is therefore equivalent to equation (53), except for the proportionality constant $3\Omega H^2/2$ in both constraint kernel \hat{H}_j and value c_j . From the above equation we can also easily infer the relation between the Fourier components $\hat{\phi}_G(\mathbf{k})$,

$$\phi_G(\mathbf{x}) = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{\phi}_G(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}}, \quad (70)$$

and the Fourier components $\hat{f}(\mathbf{k})$ of the density field,

$$\hat{\phi}_G(\mathbf{k}) = -\frac{3}{2} \Omega H^2 a^2 \frac{1}{k^2} \hat{f}(\mathbf{k}) \hat{W}^*(\mathbf{k}). \quad (71)$$

The first 3 constraints on the gravitational field therefore concern the peculiar gravitational acceleration at the position of the peak itself, $\mathbf{g}_G(\mathbf{x}_d)$. It is useful to specify it in units of the dispersion of the gravitational acceleration of peaks, $\sigma_{g,pk}(R_G) = \langle \mathbf{g}_{G,pk} \cdot \mathbf{g}_{G,pk} \rangle$,

$$g_{G,l}(\mathbf{x}_d) = \tilde{g}_l \sigma_{g,pk}(R_G), \quad l = 1, \dots, 3. \quad (72)$$

The dispersion of the peak accelerations is less than the overall dispersion σ_g of the acceleration in the field. This lowering of the acceleration of peaks compared with that of field points is caused by the extra acceleration associated with the infall of field points onto the peaks. We can infer that (see section 4.4, eqns. 101 and 106)

$$\sigma_{g,pk} = \tilde{\sigma}_g \equiv \sigma_g \sqrt{1 - \gamma_v^2}, \quad (73)$$

where $\sigma_g(R_G)$ and γ_v are given by

$$\sigma_g(R_G) = \frac{3}{2} \Omega H^2 \sigma_{-1}(R_G) \quad \text{and} \quad \gamma_v \equiv \frac{\sigma_0^2}{\sigma_{-1} \sigma_1}, \quad (74)$$

with $\sigma_j(R_G)$ the spectral moments,

$$\sigma_j^2(R_G) \equiv \int \frac{d\mathbf{k}}{(2\pi)^3} P(\mathbf{k}) \hat{W}(\mathbf{k}) k^{2j}. \quad (75)$$

The Fourier expressions for the 3 components of the bulk peculiar acceleration $\mathbf{g}_G(\mathbf{x}_d)$ can be derived from equation (65) and (70),

$$g_{G,l}(\mathbf{x}_d) = -\frac{1}{a} \frac{\partial \phi_G}{\partial x_l} = \int \frac{d\mathbf{k}}{(2\pi)^3} \frac{1}{a} i k_l \hat{\phi}_G(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}}. \quad (76)$$

Inserting equation (71) and (72) leads to the constraint equations,

$$\int \frac{d\mathbf{k}}{(2\pi)^3} \hat{f}(\mathbf{k}) \left\{ -\frac{3}{2} \Omega H^2 \frac{i k_l}{k^2} \hat{W}^*(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}_d} \right\} = \tilde{g}_l \frac{3}{2} \Omega H^2 \sqrt{1 - \gamma_v^2} \sigma_{-1}(R_G), \quad (77)$$

From this we find the corresponding constraint kernels $\hat{H}_{11}(\mathbf{k})$, $\hat{H}_{12}(\mathbf{k})$ and $\hat{H}_{13}(\mathbf{k})$, and the constraint values c_{11} , c_{12} and c_{13} ,

$$\hat{H}_j(\mathbf{k}) = \frac{3}{2} \Omega H^2 \frac{i k_l}{k^2} \hat{W}(\mathbf{k}) e^{i\mathbf{k} \cdot \mathbf{x}_d}, \quad c_j = \tilde{g}_l \frac{3}{2} \Omega H^2 \sqrt{1 - \gamma_v^2} \sigma_{-1}(R_G), \quad (78)$$

with $j = 11, \dots, 13$ and $l = 1, \dots, 3$ (see appendix F). Evidently, instead of specifying the constraint values c_j as \tilde{g}_l it is also possible to do this directly in the appropriate physical units (*e.g.* km/s²).

Five additional constraints are needed to characterize the tidal field around the peak. This field is described by the traceless (comoving) tidal tensor $E_{G,ij}$ (eq. 68). Within an arbitrary system of reference this tidal tensor is most conveniently expressed in terms of its eigenvalues and units vectors,

$$E_{G,ij} = \frac{1}{a^2} \left\{ \frac{\partial^2 \phi_G}{\partial x_i \partial x_j} - \frac{1}{3} \nabla^2 \phi_G \delta_{ij} \right\} = \sum_{k=1}^3 \mathcal{E}_k T_{ki} T_{kj}, \quad i, j = 1, 2, 3. \quad (79)$$

The elements of the matrix T_{kl} are the components of the various eigenvectors of the tidal tensor, whose directions are characterized by the 3 Euler angles α_E , β_E , and ψ_E (T_{kl} are given by equation (60), with α_E , β_E and ψ_E replacing α , β and ψ). In an initial random density field there is a strong correlation between the tidal tensor and the mass tensor $\zeta_{ij} = \nabla_i \nabla_j f$ (see section 4.4, eq. 106). In the case of peaks this translates into a strong tendency of the tidal tensor to align itself along the principal axes of the mass ellipsoid. In the specification of the initial tidal field it is therefore often useful, and physically sensible, to express its elements with respect to the reference system defined by these axes. We denote the corresponding transformation matrix by \tilde{T}_{kl} , which is defined through equation (60) by the 3 corresponding Euler angles $\tilde{\alpha}_E$, $\tilde{\beta}_E$ and $\tilde{\psi}_E$. If the orientation of the peak itself with respect to an arbitrary reference system is specified by the transformation matrix $A_{kl}(\alpha, \beta, \psi)$ (see eq. 60), then the tidal field's transformation matrix T within the same system is the matrix product of \tilde{T} with A ,

$$T_{ki} = \sum_{m=1}^3 \tilde{T}_{km} A_{mi}, \quad (80)$$

The magnitude of the tidal field in the directions of the principal axes of the tidal tensor is given by the eigenvalues \mathcal{E}_1 , \mathcal{E}_2 and \mathcal{E}_3 . Because $E_{G,ij}$ is traceless, *i.e.* $\sum \mathcal{E}_k = 0$, it is sufficient to specify two eigenvalues. To get an idea of the right order of magnitude it is usually useful to specify \mathcal{E}_k in units of σ_E , the dispersion of the off-diagonal elements of the tidal tensor $E_{G,ij}$ (see section 4.4, eq. 101),

$$\sigma_E(R_G) = \frac{3}{2} \Omega H^2 \sigma_0(R_G) \sqrt{\frac{1 - \gamma^2}{15}} \quad \text{with} \quad \gamma \equiv \frac{\sigma_1^2}{\sigma_0 \sigma_2}, \quad (81)$$

so that $\mathcal{E}_k = \tilde{\mathcal{E}}_k \sigma_E(R_G)$. An elegant and convenient parameterization of the diagonalized $E_{G,ij}$ in terms of two quantities ϵ and ϖ was introduced by Bertschinger & Jain (1994),

$$E_{G,ij} = \text{diag} [\mathcal{E}_1, \mathcal{E}_2, \mathcal{E}_3] \equiv \Omega H^2 \epsilon (1 + f_G) Q_{ij}(\varpi), \quad (82)$$

with the one-parameter traceless matrix Q_{ij} defined by

$$Q_{ij}(\varpi) \equiv \text{diag} [Q_1, Q_2, Q_3] \equiv \text{diag} \left[\cos \left(\frac{\varpi + 2\pi}{3} \right), \cos \left(\frac{\varpi - 2\pi}{3} \right), \cos \left(\frac{\varpi}{3} \right) \right]. \quad (83)$$

This matrix representation turns out to be useful when considering the Lagrangian equations of motion of a patch of matter (Bertschinger & Jain 1994). It is particularly convenient because all possible eigenvalues of $E_{G,ij}$ are obtained by $q Q_{ij}(\alpha)$, with $q \in [0, \infty)$ determining the magnitude of the tidal field, and $\alpha \in [0, \pi]$ the relative strength of the tidal field along the three principal axes. The 5 constraints, for $E_{G,11}$, $E_{G,22}$, $E_{G,12}$, $E_{G,13}$ and $E_{G,23}$, therefore have the form

$$E_{G,ij} = \tilde{\epsilon} \sigma_E(R_G) \sum_{k=1}^3 \mathcal{Q}_k(\varpi) T_{ki} T_{kj}, \quad (84)$$

with $(i, j) = (1, 1), (2, 2), (1, 2), (1, 3)$ and $(2, 3)$. Note that here we have expressed ϵ in units of σ_E , i.e. $\tilde{\mathcal{E}} = \tilde{\epsilon} \mathcal{Q}(\varpi)$. In addition, we have assumed that the fluctuations are linear, so that the factor ϵf_G can be neglected. For the generation of initial conditions, our primary interest, this assumption is not a serious restriction.

The Fourier components $\hat{E}_{G,ij}(\mathbf{k})$ of the tidal tensor,

$$E_{G,ij} = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{E}_{G,ij}(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}_d}, \quad (85)$$

can be easily found from the definition of $E_{G,ij}$ in equation (68) and subsequent differentiation and insertion of equation (71),

$$\hat{E}_{G,ij}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \left(\frac{k_i k_j}{k^2} - \frac{1}{3} \delta_{ij} \right) \hat{W}^*(\mathbf{k}) \hat{f}(\mathbf{k}). \quad (86)$$

This leads to the following tidal field constraint expressions:

$$\int \frac{d\mathbf{k}}{(2\pi)^3} \left\{ \frac{3}{2} \Omega H^2 \left(\frac{k_i k_j}{k^2} - \frac{1}{3} \delta_{ij} \right) \hat{W}^*(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}_d} \right\} \hat{f}(\mathbf{k}) = \tilde{\epsilon} \sigma_E(R_G) \sum_{k=1}^3 \mathcal{Q}_k(\varpi) T_{ki} T_{kj}, \quad (87)$$

which yields the corresponding 5 constraint kernels and values,

$$\hat{H}_l(\mathbf{k}) = \frac{3}{2} \Omega H^2 \left(\frac{k_i k_j}{k^2} - \frac{1}{3} \delta_{ij} \right) \hat{W}(\mathbf{k}) e^{i\mathbf{k} \cdot \mathbf{x}_d}, \quad c_l = \tilde{\epsilon} \frac{3}{2} \Omega H^2 \sigma_0(R_G) \sqrt{\frac{1-\gamma^2}{15}} \sum_{k=1}^3 \mathcal{Q}_k(\varpi) T_{ki} T_{kj}, \quad (88)$$

with $l = 13, \dots, 18$ and $(i, j) = (1, 1), (2, 2), (1, 2), (1, 3)$ and $(2, 3)$. Alternatively, instead of expressing the tidal constraints via the two quantities ϵ and ϖ (or \mathcal{E}) and the 3 Euler angles α_E , β_E and ψ_E , we can evidently specify the values for $E_{G,ij}$ directly, either in corresponding physical units or in units of $\sigma_E(R_G)$,

$$E_{G,ij} = \tilde{\epsilon}_{ij} \frac{3}{2} \Omega H^2 \sigma_0(R_G) \sqrt{\frac{1-\gamma^2}{15}}. \quad (89)$$

In the linear regime an analogous, and for some more familiar, way of describing the dynamics of a patch of matter is in terms of the peculiar velocity. This is possible because for growing mode linear perturbations the peculiar velocity \mathbf{v} is directly proportional to the peculiar gravitational acceleration \mathbf{g} (Peebles 1980),

$$\mathbf{v}(\mathbf{x}, t) = \frac{2\mathcal{F}(\Omega)}{3H\Omega} \mathbf{g}(\mathbf{x}, t), \quad (90)$$

where $\mathcal{F}(\Omega) \approx \Omega^{0.6}$. It is convenient to write the smoothed peculiar velocity \mathbf{v}_G around the position \mathbf{x}_d of the peak in terms of the bulk motion $\mathbf{v}_G(\mathbf{x}_d)$, the divergence $\nabla \cdot \mathbf{v}_G/a$, the shear σ_{ij} and the vorticity ω_{ij} ,

$$v_{G,i}(\mathbf{x}) = v_{G,i}(\mathbf{x}_d) + a \sum_{j=1}^3 \left\{ \frac{1}{3a} (\nabla \cdot \mathbf{v}_G)(\mathbf{x}_d) \delta_{ij} + \sigma_{ij}(\mathbf{x}_d) + \omega_{ij}(\mathbf{x}_d) \right\} (x_j - x_{d,j}). \quad (91)$$

The shear is the trace-free symmetric part of $\partial v_{G,i}/\partial r_j$,

$$\sigma_{ij} = \frac{1}{2a} \left\{ \frac{\partial v_{G,i}}{\partial x_j} + \frac{\partial v_{G,j}}{\partial x_i} \right\} - \frac{1}{3a} (\nabla \cdot \mathbf{v}_G) \delta_{ij}, \quad (92)$$

while the vorticity ω_{ij} is the antisymmetric part,

$$\omega_{ij} = \frac{1}{2a} \left\{ \frac{\partial v_{G,i}}{\partial x_j} - \frac{\partial v_{G,j}}{\partial x_i} \right\}. \quad (93)$$

Because ω_{ij} does not have a gravitational origin, it is an irrelevant quantity as far as constraints on the density perturbation field are concerned. Moreover, it can be shown to remain zero whenever there is no primordial vorticity (Peebles 1980). From this we can infer that constraints on the peak velocity $\mathbf{v}_G(\mathbf{x}_d)$ are therefore equivalent to equation (77), except that the factor $3\Omega H^2 a/2$ in both the constraint kernel \hat{H}_j and constraint value c_j has to be changed into $Ha\mathcal{F}(\Omega)$. Also, the constraint on the divergence $\nabla \cdot \mathbf{v}_G/a$ is equivalent to the constraint on $f_G(\mathbf{x})$ (eq. 53), except for a factor $H\mathcal{F}(\Omega)$. Finally, we see that the relation between the shear $\sigma_{G,ij}$ and $E_{G,ij}$ is

$$E_{ij} = -\frac{3}{2}\Omega H^2 \frac{\sigma_{ij}}{H\mathcal{F}(\Omega)}. \quad (94)$$

We should, however, not fail to appreciate that in the nonlinear regime the simple relation between \mathbf{v} and \mathbf{g} breaks down. In that case the same basic physical relationships between the acceleration \mathbf{g} , the potential ϕ and the density ρ remain valid. As this is not true for the velocity \mathbf{v} , it is fundamentally preferable to impose constraints on the gravitational field instead of the velocity field.

4.4 Probability of peak constraints

In principle, in the case of a Gaussian field any set of numerical values for the 18 peak constraints per peak is possible, regardless of how small the probability of the occurrence of such peaks would be. This is a consequence of the ability of the Hoffman-Ribak constrained random field method to generate realizations for *any arbitrary* set of values for the imposed constraints. In order to prevent the generation of unlikely circumstances it is therefore necessary to control or have an estimate of the likelihood of the constraints. The corresponding probability distribution of the constraints is given by equation (15). A good measure of the likelihood can be obtained by calculating the χ^2 ,

$$\chi^2 = \sum_{i,j=1}^M C_i (\mathbf{Q}^{-1})_{ij} C_j = c_i \xi_{ij}^{-1} c_j. \quad (95)$$

The probability that for this constraint set χ^2 has this value or higher can then be directly calculated from $\Gamma_Q(M/2, \chi^2/2)$, where Γ_Q is the incomplete gamma function. As a rule of a thumb, the constraint set can be considered to represent manifest unlikely conditions, if the χ^2 per degree of freedom, $\tilde{\chi}^2 \equiv \chi^2/M$, differs significantly from unity. Note that the computational cost of evaluating $\tilde{\chi}^2$ is negligible as the inverse of the constraint-constraint correlation matrix $\xi_{ij} = \langle C_i C_j \rangle$ has already been calculated as part of the construction procedure (eq. 40 and 44).

A full expression for χ^2 or, even better, the full probability distribution in terms of the 18 constraint quantities can be obtained by evaluating the expression in equation (15), following the treatment presented in Appendix A of BBKS. Following the discussion in the previous sections the density and gravity field in and around an arbitrary point \mathbf{x} in a Gaussian random density field $f(\mathbf{x})$ can be characterized by 18 parameters

$$\Upsilon = \{\nu, \eta_1, \eta_2, \eta_3, \zeta_1, \zeta_2, \zeta_3, \zeta_4, \zeta_5, \zeta_6, g_1, g_2, g_3, E_1, E_2, E_4, E_5, E_6\}, \quad (96)$$

with $f = \nu\sigma_0$ the value of the field at \mathbf{x} , $\nabla_i f = \eta_i$ the first derivatives of the field, and ζ_A the six independent components of the tensor $\zeta_{ij} = \nabla_i \nabla_j f$ (where $A = 1, 2, 3, 4, 5, 6$ refer to the $ij = 11, 22, 33, 12, 13, 23$ components of the tensor). In addition, $g_i = -\nabla_i \phi$ is the peculiar gravitational acceleration, while E_A are the five independent components of the traceless tidal tensor $E_{ij} = \nabla_i \nabla_j \phi - \frac{1}{3} \nabla^2 \phi \delta_{ij}$ (with $A = 1, 2, 3, 4, 5, 6$ referring to the $ij = 11, 22, 33, 12, 13, 23$ components of E_{ij}).

The probability $\mathcal{P}(\Upsilon)$ that at the position \mathbf{x} the field has the specified values for these 18 quantities is specified by a joint Gaussian probability distribution, for which a reasonably insightful expression can be found by reducing the corresponding 18×18 covariance matrix $\mathbf{Q} = \langle y_i y_j \rangle$ into a block diagonal matrix of 9 2×2 blocks. This is achieved by transformation of the set of variables $\{\zeta_1, \zeta_2, \zeta_3, E_1, E_2\}$ into a new set $\{x, y, z, E_y, E_z\}$,

$$\begin{aligned} x &= -\frac{\zeta_1 + \zeta_2 + \zeta_3}{\sigma_2}, & y &= -\frac{\zeta_1 - \zeta_3}{2\sigma_2}, & z &= -\frac{\zeta_1 - 2\zeta_2 + \zeta_3}{2\sigma_2}, \\ E_y &= \frac{E_1 - E_3}{2}, & E_z &= \frac{E_1 - 2E_2 + E_3}{2}. \end{aligned} \quad (97)$$

$\mathcal{P}(\Upsilon)$ is then given by

$$\mathcal{P}(\Upsilon) = A e^{-Q/2} d\nu d^3\eta dx dy dz d\zeta_4 d\zeta_5 d\zeta_6 d^3g dE_y dE_z dE_4 dE_5 dE_6, \quad (98)$$

with

$$A = \frac{3^6 5^5}{1024 \pi^9 (1 - \gamma^2)^3 (1 - \gamma_v^2)^{3/2} (\frac{3}{2} \Omega H^2)^4 \sigma_{-1}^3 \sigma_0 \sigma_1^3 \sigma_2^3}, \quad (99)$$

and

$$\begin{aligned}
Q &= \sum_{i,j=1}^{18} y_i (\mathbf{Q}^{-1})_{ij} y_j = \\
&= \nu^2 + \frac{(x - x_*)^2}{1 - \gamma^2} + 15y^2 + 5z^2 + \frac{3\vec{\eta} \cdot \vec{\eta}}{\sigma_1^2} + \sum_{A=4}^6 \frac{15\zeta_A^2}{\sigma_2^2} + \\
&\quad \frac{3(\vec{g} - \vec{g}_*)^2}{\tilde{\sigma}_g^2} + \frac{(E_y - E_y^*)^2}{\sigma_E^2} + \frac{(E - E_z^*)^2}{3\sigma_E^2} + \sum_{A=4}^6 \frac{(E_A - E_A^*)^2}{\sigma_E^2},
\end{aligned} \tag{100}$$

where $\tilde{\sigma}_g$ and σ_E are defined by

$$\tilde{\sigma}_g \equiv \left(\frac{3}{2}\Omega H^2\right) \sigma_{-1} \sqrt{1 - \gamma_v^2}, \quad \sigma_E \equiv \left(\frac{3}{2}\Omega H^2\right) \sigma_0 \sqrt{\frac{1 - \gamma^2}{15}}, \tag{101}$$

while the various coupling quantities x_* , \vec{g}_* , E_y^* , E_z^* and E_A^* are defined by

$$\begin{aligned}
x_* &= \gamma \nu, \\
\vec{g}_* &= \gamma_v \left(\frac{3}{2}\Omega H^2\right) \frac{\sigma_{-1}}{\sigma_1} \vec{\eta} \\
E_y^* &= \gamma y \left(\frac{3}{2}\Omega H^2\right) \sigma_0, \quad E_z^* = \gamma z \left(\frac{3}{2}\Omega H^2\right) \sigma_0, \\
E_A^* &= \gamma \left(\frac{3}{2}\Omega H^2\right) \frac{\sigma_0}{\sigma_2} \zeta_A, \quad A = 4, 5, 6.
\end{aligned} \tag{102}$$

(for the definitions of γ , γ_v and the various σ_j see eqns. 81, 74 and 75). In the case of a peak we can further reduce the expression for Q . Evidently, $\vec{\eta} = 0$. In addition, we can use the fact that Q should be independent of the orientation of the mass ellipsoid around the peak, expressed by its Euler angles α , β and ψ . We can therefore discard the orientation term $\sum 15\zeta_A^2/\sigma_2^2$ and redefine x , y and z in terms of the eigenvalues λ_i of $(-\zeta_{ij})$, whose relation to the axis ratios of the ellipsoid are given in equation (62),

$$x = \frac{\lambda_1 + \lambda_2 + \lambda_3}{\sigma_2}, \quad y = \frac{\lambda_1 - \lambda_3}{2\sigma_2}, \quad z = \frac{\lambda_1 - 2\lambda_2 + \lambda_3}{2\sigma_2}, \tag{103}$$

Furthermore, we restrict the ordering of the eigenvalues to $\lambda_1 \geq \lambda_2 \geq \lambda_3 > 0$. The condition that $\lambda_3 > 0$ is equivalent to the demand that ζ_{ij} has to be negative definite, which, together with the constraints $\nabla_i f$, is necessary and sufficient to have a local density maximum, a peak. We then find for the complete probability $\mathcal{P}(\Upsilon)$ that at an arbitrary position \mathbf{x} there is a peak with a height $f = \nu\sigma_0$, a shape characterized by the parameters x , y and z , an orientation specified by the Euler angles α , β and ψ , an acceleration \vec{g} , and a tidal field described by the parameters E_y , E_z , E_4 , E_5 and E_6 , or, rather, that there is a peak with these parameters in the specific infinitesimal ranges around these values,

$$\begin{aligned}
&\mathcal{P}(\nu, x, y, z, \alpha, \beta, \psi, \vec{g}, E_y, E_z, E_4, E_5, E_6) \\
&= \tilde{A} |y(y^2 - z^2)| \sin \beta e^{-\tilde{Q}/2} d\nu d^3\eta dx dy dz d\alpha d\beta d\psi d\vec{g} dE_y dE_z dE_4 dE_5 dE_6.
\end{aligned} \tag{104}$$

In this equation the constant \tilde{A} is

$$\tilde{A} = \frac{3^7 5^5}{256 \pi^9 (1 - \gamma^2)^3 (1 - \gamma_v^2)^{3/2} (\frac{3}{2} \Omega H^2)^4 \sigma_{-1}^3 \sigma_0 \sigma_1^3}, \quad (105)$$

and

$$\tilde{Q} = \nu^2 + \frac{(x - x_*)^2}{1 - \gamma^2} + 5(3y^2 + z^2) + \frac{3\vec{g}^2}{\vec{\sigma}_g^2} + \frac{(E_y - E_y^*)^2}{\sigma_E^2} + \frac{(E_z - E_z^*)^2}{3\sigma_E^2} + \sum_{A=4}^6 \frac{E_A^2}{\sigma_E^2}, \quad (106)$$

where E_y , E_z , E_4 , E_5 and E_6 are specified with respect to the principal axis of the mass ellipsoid (see section 4.3, eq. 80 for the appropriate transformations). In particular, this implies that $E_A^* = 0$ for $A = 4, 5, 6$ (eq. 102). Also note that equation (100) is therefore an expression of the fact that in an initial random density field the tidal field has a strong preference to align itself along the principal axes of the the mass tensor ζ_{ij} . In particular, for a peak this implies that the strongest tidal force tends to be directed along its smallest axis (the one with the highest eigenvalue λ_i). As it explicitly takes into account this strong correlation between the initial mass quadrupole and the tidal field at the position of the peak, the reference system defined by the mass ellipsoid is therefore the most natural one to specify the initial tidal forces. The resulting expression for \tilde{Q} in the above equation is essentially the one for the χ^2 of the imposed constraints, once it is scaled to the appropriate filter and filter scale R_G by means of $\gamma(R_G)$, $\gamma_v(R_G)$ and the various spectral moments $\sigma_j(R_G)$.

The probability distribution in equation (104) is the one for having, at some arbitrary field position, a peak with the required physical parameters. Often, however, we are more interested in the more specific question what the probability \mathcal{P}_{pk} is that a peak at an arbitrary position has these imposed constrained properties. To evaluate this we need to determine the (comoving) number density of peaks with the constrained parameters, which can be done following the prescription in BBKS. To obtain \mathcal{P}_{pk} this specific number density has to be divided by the total comoving number density of peaks, n_{pk} , whose value is (see BBKS),

$$n_{pk} = \frac{29 - 6\sqrt{6}}{5^{3/2} 2(2\pi)^2 R_*^3} = 0.016 R_*^{-3}, \quad \text{with} \quad R_* \equiv \sqrt{3} \frac{\sigma_1}{\sigma_2}. \quad (107)$$

From this we can derive the probability that a peak has a height $\nu\sigma_0$, shape parameters x , y and z , an acceleration \vec{g} and tidal tensor components E_y , E_z , E_4 , E_5 and E_6 . Since the orientation of the peak is here a less relevant quantity we integrate over the Euler angles α , β and ψ . This can be done without further complications since \tilde{Q} is independent of these Euler angles. Note that this automatically implies that the tidal tensor components are specified with respect to the principal axes of the mass ellipsoid. We then obtain the following expression for \mathcal{P}_{pk} ,

$$\begin{aligned} P_{pk}(\nu, x, y, z, \vec{g}, E_y, E_z, E_4, E_5, E_6) d\nu dx dy dz d\vec{g} dE_y dE_z dE_4 dE_5 dE_6 \\ = \tilde{B} \Theta(x, y, z) F(x, y, z) e^{-\tilde{Q}/2} d\nu dx dy dz d\vec{g} dE_y dE_z dE_4 dE_5 dE_6. \end{aligned} \quad (108)$$

In this expression \tilde{Q} is given by equation (106), while

$$F(x, y, z) = y(y^2 - z^2)(x - 2z)[(x + z)^2 - (3y)^2]. \quad (109)$$

In addition, the function $\Theta(x, y, z)$ is defined such that its value is 1 when the peak constraints in the (x, y, z) domain are satisfied, and 0 otherwise. These constraints are that $y \geq z \geq -y$, $y \geq 0$, to obtain the correct ordering of the eigenvalues λ_i of ζ_{ij} , and $(x + z - 3y) > 0$ so that the smallest eigenvalue λ_3 is positive and we indeed have a peak. The constant \tilde{B} is given by

$$\tilde{B} = \frac{5^{13/2} 3^{17/2}}{16 \pi^5 (29 - 6\sqrt{6}) (1 - \gamma^2)^3 (1 - \gamma_v^2)^{3/2} (\frac{3}{2} \Omega H^2)^4 \sigma_{-1}^3 \sigma_0}. \quad (110)$$

This can be easily extended to other conditional peak probabilities, e.g. the chance that a peak of height $\nu\sigma_0$ has the required parameters. However, calculating the involved expressions quickly becomes a very elaborate procedure.

In the above we have mainly concentrated on the probability of constraints imposed on one particular peak, with the intention of providing insight into how the different constraints interrelate and to get an idea of the expected order of magnitude of each of the constraints. However, as we have seen earlier, our code allows to provide constraints on many different peaks, at different positions and scales. Giving analytical expressions for such constraints would quickly become a cumbersome and elaborate affair, due to the introduction of spatial correlations in the random field. However, via equation (95) the numerical value of χ^2 for these constraints can be easily computed, providing a good idea of their likelihood.

5. Realizations and Applications

The formalism developed in the previous sections allows the generation of a large variety of initial conditions. In this section we will visualize the procedure by providing some practical examples. The versatile and non-local nature of the formalism has already been emphasized in figure 1 (section 2), illustrating the construction of a density field that is constrained to have two peaks of a different scale, a different shape, and at different positions. Although for the construction of the density field around one central peak a few other equally or even more efficient methods have been developed (Binney & Quinn 1991, Bond & Meyers 1993), mainly based on a multipole expansion of the field, their efficiency breaks down if the constraints are imposed at more than one position.

5.1 Field-constraint correlations

At the core of our construction procedure is the superposition of a mean field \bar{f} and a properly sampled residual field F (see eq. 26). The mean field \bar{f} is effectively the superposition of the field-constraint correlation fields $\xi_i(\mathbf{x})$ (eq. 22), each weighted by a factor $\sum_j \xi_{ij}^{-1} c_j$.

Figure 3 shows the mean field and five of the composite field-constraint correlation fields for the set of peak constraints described below. These realizations are generated in a periodic $100h^{-1}$ Mpc box. As for figure 1, the power spectrum of the random field fluctuations is the standard cold dark matter spectrum of Davis et al. (1985) with $\Omega_{\text{CDM}} = 1.0$ and $h = 0.5$ — normalized such that $\sigma(8h^{-1} \text{ Mpc}) = 1.0$ at $a = 1$, the present epoch (Davis & Peebles 1983). The panels in figure 3

Figure 3. The mean field \bar{f} (top left panel) and five of the composite field-constraint correlation fields $\xi_i(\mathbf{r})$ of a set of constraints (see text). The panels contain the contourmaps of the density (top left, contour spacing 0.5) and correlation values (other 5 panels) in a $5h^{-1}$ Mpc slice through the center of a $100h^{-1}$ Mpc box. The spectrum of the field is the standard CDM spectrum. The 5 field-constraint correlation functions are a) top middle: $\langle f f_G \rangle$ (contour spacing 0.1), b) top right: $\langle f \nabla_x f_G \rangle$ (contour spacing 0.03) c) bottom left: $\langle f \nabla_x^2 f_G \rangle$ (contour spacing 0.02), d) bottom middle: $\langle f v_{G,x} \rangle$ (contour spacing 0.03) and e) $\langle f \sigma_{xx} \rangle$ (contour spacing 0.015), with f_G the value of the smooth density field, $v_{G,x}$ the x -component of the peculiar velocity and σ_{xx} the xx -component of the shear tensor, all evaluated at the center of the box.

contain contourmaps of the density and correlation values in a $5h^{-1}$ Mpc slice centered halfway in the simulation box, each of the maps being smoothed on a Gaussian scale of $4h^{-1}$ Mpc.

All the constraints are defined on a Gaussian scale of $4h^{-1}$ Mpc. A triaxial peak, with axis ratio $10 : 9 : 7$, of height $f_G = 3\sigma_0$ and local density field curvature $\nabla^2 f_G = \langle x \rangle \sigma_2 \approx 3.481\sigma_2$ is positioned at the center of the box. Its major axes are slightly oriented with respect to the coordinate axes of the box. In addition to these local constraints, there are constraints on the local gravity and tidal field. Because we limit ourselves to growing mode linear perturbations we specify these constraints in terms of the peculiar velocity and shear. The total peculiar velocity of the peak is 1145 km/s , towards a direction 26.6° “north” of the positive x -axis and 22.6° out of the $x - y$ plane, in the positive z -direction (note that the specified numerical values of the constraints are the linear extrapolations to the present epoch, $a = 1$). This corresponds to a value of 2.00 times the velocity dispersion of peaks on a scale of $4h^{-1}$ Mpc, or 1.66 times the velocity dispersion of an average field point on this scale (the lower value of peak velocities in comparison with the velocity of field points is due to the extra component corresponding to accretion onto the peaks). The shear tensor at the location of the peak is orientated so that the off-diagonal terms are zero. The diagonal term in the x -direction, σ_{xx} , has the largest magnitude and is positive (dilation), while σ_{yy} and σ_{zz} are equal and negative (contraction). For illustrative purposes we have chosen a rather extreme value for the magnitude of the largest element of the shear tensor: 100 km/s/Mpc on the scale of $4.0h^{-1}$ Mpc, ≈ 6.8 times the dispersion ($\approx 14.5 \text{ km/s/Mpc}$) for the diagonal shear components for peaks (Bond 1987). A good idea of the order of magnitude of these shear tensor values is obtained by comparison with the value of the expansion scalar for the $3\sigma_0$ peak, $\nabla \cdot \mathbf{v}_G = -142.47 \text{ km/s}$.

The specified constraints can be easily recognized in the resulting mean field, in the top left panel of figure 3 (contour spacing 0.5, and the positive value solid contours separated from the negative value dotted contours by the thick solid line corresponding to $f = 0$). The contours around the center of the box clearly reveal the presence of the elongated peak, oriented with respect to the coordinate axes. The global density field in the box reflects the gravity and tidal field constraints. The source of the motion of the peak is the concentration of mass in the upper righthand quarter of the frame, while the clearly discernable quadrupolar component in the matter distribution induces the tidal field. To understand how the different constraints conspire to produce this mean field it is quite revealing to study the individual field-constraint correlation functions $\xi_i(\mathbf{x})$. The five illustrated correlation functions $\xi_i(\mathbf{x})$ are the correlation of the field $f(\mathbf{x})$ with (1) the value of the smoothed density at the peak position \mathbf{x}_i , $f_G(\mathbf{x}_i)$ (top middle panel), (2) the value of the first derivative of the smoothed density field $\nabla_x f_G(\mathbf{x}_i)$ (top right panel), (3) the value of the second derivative of the smoothed density field $\nabla_x^2 f_G(\mathbf{x}_i)$ (bottom left panel), (4) the peculiar velocity $v_{G,x}(\mathbf{x}_i)$ (bottom middle panel) and (5) the shear component $\sigma_{xx}(\mathbf{x}_i)$ (bottom right panel).

Figure 4. The variance of constrained random field realizations. The mean field \bar{f} and four different field realizations of a set of constraints (see figure 3) are shown in the top left panel and the middle and right rows respectively. The panels contain the density contourmaps in the $5h^{-1}$ Mpc thick central slice in a $100h^{-1}$ Mpc box. The contour spacing is 0.5. The bottom left panel is the contourmap of the value of the variance of the field realizations inside the slice, running from 0.0 at the centre to $\sigma_0 \approx 0.95$ at the edge of the box (contour spacing 0.05).

The first correlation function (top middle panel) is spherically symmetric, with a value of 1.0 near the centre, and radially decreasing to a value of 0.0 at the outer contour (contour spacing is 0.1). Effectively, this correlation function is the convolution between the field correlation function $\xi(\mathbf{x}) = \langle ff \rangle$ and the Gaussian filter function defining the scale of the constrained object. In a similar fashion we can consider the second correlation function (top right panel, contour spacing 0.03) to be the convolution of the correlation function $\xi(\mathbf{x})$ with the first derivative of the filter function. This introduces the anisotropy along the x -axis, with, within distances comparable to the correlation radius, negative values on the left side of the peak and positive values to the right. Further outward the correlation function $\xi(\mathbf{x})$ becomes negative, resulting in the sign reversal of $\xi_i(\mathbf{x})$. The third correlation function (bottom left panel, contour spacing 0.02) is essentially the convolution of the field correlation function $\xi(\mathbf{x})$ with the second derivative of the Gaussian function, $\partial^2 W_G / \partial^2 x$. Because this derivative has two zero-points along the x -axis we see a negative value near the centre, changing to positive on both sides of the centre.

The correlation functions corresponding to the velocity and shear constraints display familiar patterns. The function corresponding to the constraint on the peculiar velocity in the x -direction, v_x , (middle bottom panel, contour spacing 0.03) is a dipolar function centred on the position of the peak, with positive values to the righthand side of the peak (in the x -direction) and negative values to the left. This is evidently related to the fact that such a dipolar matter distribution would produce a net gravitational acceleration, and corresponding peculiar velocity, in the x -direction. In addition, we see that the constraint on the shear component σ_{xx} results in a clear quadrupolar pattern of the correlation function (right bottom panel, contour spacing 0.015). As in the case of the velocity-field correlation function, this is related to the non-zero tidal force xx -component that would be produced by such a quadrupolar mass distribution.

Superposition of the complete set of these correlation fields $\xi_i(\mathbf{x})$, with the appropriate weight factors, proportional to the corresponding constraint values c_i , produces the mean field in the top left panel. Comparison of the different panels in figure 4 shows that several of the correlation function patterns can indeed be recognized in the mean field.

5.2 Variance of realizations

After having constructed the mean field \bar{f} (former subsection), we need to add a properly sampled residual field realization (eq. 26). Figure 4 provides an idea of the possible variations between the residual field realizations and, specifically, the resulting full field realizations. In addition to the mean field illustrated in the the top left panel, four different realizations are shown in the middle and right row of panels. All these panels are density contour maps (contour spacing 0.5) in the same central $5h^{-1}$ Mpc thick slice used in figure 3. From the four field realizations we can infer that, for example, the mass concentration to the right, responsible for the peculiar motion of the peak, can vary substantially in position, shape, size and substructure. Moreover, the morphology

and distribution of mass clumps inside the band of matter along the x -axis, main contributor to the specified shear, displays an even larger variation, in particular at large distances from the peak.

An analytic expression for the variance of the residual field at any position \mathbf{x} follows immediately from the independence of the residual field distribution function from the numerical values of the imposed constraints (eq. 32), see Appendix D,

$$\langle F^2(\mathbf{x})|\Gamma \rangle = \sigma_0^2 - \xi_i(\mathbf{x})\xi_{ij}^{-1}\xi_j(\mathbf{x}), \quad (111)$$

with

$$\sigma_0^2 = \langle f^2(\mathbf{x}) \rangle, \quad (112)$$

the variance of the density field (recall that both f and F have zero mean). The expression in equation (111) shows that $\langle F^2(\mathbf{x})|\Gamma \rangle$ is dependent on \mathbf{x} , and therefore implies the residual field $\langle F^2(\mathbf{x})|\Gamma \rangle$ is neither homogeneous nor isotropic. Note that because $F(\mathbf{x})$ is a Gaussian random field its distribution functional $\mathcal{P}[F|\Gamma]$ is completely specified by the variance $\langle F^2(\mathbf{x})|\Gamma \rangle$.

The lower left panel shows a contour map of the variance field corresponding to the constraint set in the example. Notice the perfect spherical character of this variance field, increasing radially outward from the position of the peak, where it is equal to 0.0, to the general field value $\sigma_0 \approx 0.95$ (contour spacing 0.05). At first sight this might seem counterintuitive, as most of the applied constraints are non-isotropic. However, from equation (111) we see that $\langle F^2(\mathbf{x})|\Gamma \rangle$ involves a product of all field-constraint correlation functions $\xi_i(\mathbf{x})$, independent of the actual numerical values c_i of the constraints. In our example all 18 peak constraints have been specified. This means that the anisotropy introduced by e.g. the dipole distribution corresponding to the v_x constraint gets fully compensated by the equally strong y and z dipole distributions of the v_y and v_z field-constraint correlation functions. The same is true for the quadrupole distributions of the shear constraints, as well as for the correlation functions corresponding to the three first derivatives $\partial f_G/\partial x_k$ and the six second derivatives $\partial^2 f_G/\partial x_k \partial x_l$.

The predicted variance field can also be recognized when comparing the four field realizations. They show very small differences in the neighbourhood of the peak, but further outward the differences become larger and ultimately are equal to the variations in any average field.

5.3 Realizations for Gravity and Tidal Field constraints

An important ingredient of our code is the ability to put constraints on the peculiar gravity or the tidal field acting on a peak. While the peaks in figure 1 do not have constraints on either the gravity and tidal field, we intend to give an impression of the consequences for both density and velocity fields of imposing such constraints by means of a sequence of four random field realizations. Each of the four examples contain the same peak at the center of the box, but differ in the constraints on the gravity and tidal field to which the peak is subjected. The central $3\sigma_0$ peak is defined on a Gaussian scale of $4h^{-1}$ Mpc, is spherical in shape, and has a peak curvature of $\nabla^2 f_G = \langle x \rangle \sigma_2 \approx 2.901\sigma_2$. By using the same random number generator for each of the realizations we try to keep

the differences between the residual fields at a minimum (however, note from eq. 111 that there will be differences depending on which constraints are applied).

In the first example (A, figure 5a) the central peak is not subjected to any velocity and shear field constraints. In the case of the second example (B, figure 5b), the same peak is constrained to have a peculiar velocity of 1000 km/s in the positive x -direction (note that the specified numerical values of these quantities are the linear extrapolations to the present epoch, $a = 1$, and that we specify the gravity and tidal field constraints in terms of peculiar velocity and shear). In the third realization (C, figure 5c) we constrain the shear at the peak's position, while its peculiar velocity is unconstrained. The off-diagonal terms of the shear tensor are zero, while σ_{xx} has a positive value of 50 km/s/Mpc on the scale of $4.0h^{-1}$ Mpc and σ_{yy} and σ_{zz} have equal and negative values. In the final, fourth, realization (D, figure 5d) we combine the constraints to the peculiar velocity in the second example and the shear at the position of the peak in the third example.

The density and velocity field realizations for the four different constraint sets are the subject of figure 5. In all four cases we use a set of six panels to highlight different aspects of the fields, with each panel illustrating a density or velocity field in the same $5h^{-1}$ Mpc planar section along the z -direction, centered halfway in the simulation box. The different contributions to the constrained density fields are shown in the top row panels, in combination with the corresponding velocity fields in the bottom row. The top left panel contains the contourmap of the mean density field, smoothed by a Gaussian filter with a scale of $2h^{-1}$ Mpc (contour spacing is equal to $0.65=0.376\sigma_0(2h^{-1} \text{ Mpc})$). The corresponding mean peculiar velocity field is represented by the vector velocity field in the panel below. The arrows are the projections of the velocity vectors, for presentation purposes we limit ourselves to show them at the positions of the gridpoints of a 32^3 grid. The length of each arrow is proportional to the magnitude of the velocity, a length of $1/20th$ of the boxlength corresponding to a velocity of 1000 km/s. The corresponding full density field realization is represented by two panels, a density contour map of the density field (top middle panel), smoothed on the constraint scale of $4h^{-1}$ Mpc, and a Zel'dovich particle distribution (top right panel). The constraints will heavily influence the wavevectors on a scale comparable to and larger than the scale on which they are imposed, while the smaller scale waves, responsible for the subclumps and other small scale features, are not very much affected due to their negligible correlation with the imposed constraints (compare eq. 38 and the listing of constraint kernels $\hat{H}(\mathbf{k})$ in Appendix F). The contourmap in the top middle panel (contour spacing $0.275=0.290\sigma_0(4h^{-1} \text{ Mpc})$) is therefore the best illustration of that part of the density field affected by the constraints. The particle distribution, on the other hand, shows the contribution of the small scale waves to the density field at highest possible resolution. The particle positions were obtained by using the Zel'dovich approximation to evolve an initial distribution of 64^3 particles to an expansion factor $a = 0.4$, approximately the time at which the maximum density fluctuation on the scale of 1 gridcell is equal to 10.0. An additional advantage of this particle distribution is that it provides a good representation of how the density field evolves deep into the quasi-linear regime. The velocity vector field in the bottom right panel is the unsmoothed full velocity field realization, and is closely related to the Zel'dovich particle distribution. The resolution of this velocity field representation is essentially that of one gridcell

Figure 5. Four different realizations of constrained random fields in the standard cold dark matter scenario ($\Omega = 1.0, h = 0.5$). The constraints are specified on a Gaussian scale of $4h^{-1}$ Mpc. In all cases there is the same $f_G = 3\sigma_0$ spherical peak, with standard curvature $\nabla^2 f_G \approx 2.901\sigma_2$, at the center of the box. In (a) no further constraints are specified. In (b) the peak is constrained to move with a peculiar velocity of 1000 km/s towards the positive x -direction. In (c) the diagonal components of the traceless shear tensor are constrained to have the value $\sigma_{xx} = 100$ km/s/Mpc and $\sigma_{yy} = \sigma_{zz} = -50$ km/s/Mpc while the off-diagonal components are all zero. In (d) the spherical peak has the combined velocity and shear constraints of (b) and (c). The four examples are illustrated by six panels. All show an aspect of the density or velocity field in the $5h^{-1}$ Mpc thick central slice of the $100h^{-1}$ Mpc box. Top left panel: the $2h^{-1}$ Mpc smoothed density contourmap of the mean field \bar{f} , contour spacing 0.65. Top middle panel: the $4h^{-1}$ Mpc smoothed density contourmap of the constrained field realization f , contour spacing 0.275. Top right panel: Zel’dovich particle distribution at the epoch for which the maximum density fluctuation is $f = 10.0$ on the scale of one gridcell. Bottom left panel: mean velocity vector map corresponding to mean density field \bar{f} . The vectors are the projected velocity vectors in this plane. A vector with a length of $1/20$ th of the boxsize represents a velocity of 1000 km/s. All velocity vector maps were determined on a 64^3 grid, but for presentation purposes only the vectors on the gridpoints of a 32^3 subgrid are shown. Bottom middle panel: vector map of the constrained velocity field, Gaussian smoothed on a scale of $4h^{-1}$ Mpc. Bottom right panel: unsmoothed constrained velocity field vector map.

in the 64^3 grid that was used to perform the constrained field calculations. Filtering this velocity field with a Gaussian function of radius $4h^{-1}$ Mpc yields the velocity field in the bottom middle panel, corresponding to the smoothed density field in the panel above it.

A perfectly spherical density distribution around a maximum at the center of the box is evidently the mean density field in example A, with pure spherical infall characterizing the vector velocity field (left row of figure 5a). In a technical sense, recalling the discussion on figure 3, we can understand the spherical density field as the superposition of the spherical correlation function $\langle f f_G \rangle$ (top middle panel) and three equally large contributions from the correlation functions $\langle f \nabla_x^2 f_G \rangle$, $\langle f \nabla_y^2 f_G \rangle$, and $\langle f \nabla_z^2 f_G \rangle$ (bottom left panel), whose main effect is to produce a slightly flatter peak. The spherically shaped peak can also be recognized in the center of the full field realization. However, the shape of the central clump becomes very irregular further outward from the center. A comparison with the Zel’dovich particle distribution shows that this clump consists of at least four separate subclumps. Note that the central peak, unlike the peak in the mean field, has a considerable peculiar motion in the negative y direction, and a small but nonzero shear. Both are introduced via the residual field. The absence of correlations between the small scale waves is well illustrated by the full velocity field in the bottom right panel of figure 5a, which besides spherical infall does not appear to display any additional features but the expected noise.

The character of the field realization changes considerably by adding the extra constraint that the central spherical peak has a peculiar velocity of 1000 km/s in the x -direction (example B, figure 5b). The presence of the central spherical peak can still be recognized in the mean density field and the full field realization. At the same time we see that the global matter distribution is sculpted into the dipolar pattern that induces the net gravitational acceleration corresponding to the required peculiar velocity. The mean velocity field in the neighbourhood of the central density peak clearly reflects the required bulk motion. This local motion is part of a more global pattern in the velocity field, consisting of a convergence towards one point, ‘attractor’, in the right half and an outflow pattern from the underdense regions in the left half. Besides this mean component, the full velocity field realization contains additional local features, clearly visible in the lower middle and right panel of figure 5b. Note that there are several local regions from which matter is streaming away, some of these local density depressions are not even underdense (note e.g. the saddlepoint around $[x, y] = [70.0, 50.0]h^{-1}$ Mpc). Also remark the fact that the central

peak is more compact than in example A, mainly due to the very steep density falloff of the peak on the side where it is lying on the boundary of the underdense region. This pattern finds its origin in the extra superposition of the dipolar pattern characteristic of the correlation function $\langle f v_{G,x} \rangle$ (see figure 3). Equally striking are the consequences of imposing extra constraints, in example C, on the tidal field and/or corresponding shear at the peak position (figure 5c). The constraints induce the expected global quadrupolar mass distribution in the mean density field, superimposed on the local spherical peak density distribution. The band of matter parallel to the x -axis, visible in both the mean and final density field, induces the dilational shearing motion along the x -direction and the compressional shear along the other two directions, in collaboration with the underdense regions below and on top of it. The presence at the peak position of the positive σ_{xx} component, along with the negative σ_{yy} component of half its magnitude, is most strikingly visible in the mean velocity field. In the full field realization we can also recognize the presence of other components than the quadrupolar one. The central high-density ridge is littered with numerous small scale peaks of different sizes (see e.g the Zel'dovich particle distribution) while a clear dipolar component can also be discerned in the density distribution. High-density regions are concentrated in the lower half of the box, inducing the sizable peculiar motion of the peak towards the negative y -direction that can be seen in the velocity field realizations in the lower middle and right panels. Finally, figure 5d shows how the combination of the constraints on the peculiar velocity and the shear in example D work out. The corresponding mean density field clearly contains both a dipolar and a quadrupolar component, both of which are also conspicuously present in the full density field realization (also compare with the Zel'dovich particle distribution). In both the mean velocity field and the full velocity field realization we can recognize the specified peculiar velocity and shear at the position of the central peak. The particle distribution shows that the clumps on the right hand side of the center are more massive than the ones in figure 5c. The agglomerate of these clumps conspires to form a big attractor, easily recognizable, that induces the large peculiar motion of the peak.

6. Summary and Discussion

In this paper we have developed a formalism to set up cosmological initial Gaussian random density and velocity fields that can contain one or more peaks or dips, with the intention to generate appropriate initial conditions for cosmological N -body simulations that focus on the evolution of the progenitors of the present-day galaxies and clusters and their environment. The method is suited for fields with any arbitrary power spectrum $P(k)$. Central objective of our algorithm is the ability to sculpt the local and global matter distribution in a sufficiently large volume such that certain physical characteristics of the density and velocity field in the immediate neighbourhood of the primordial peaks have a priori specified values. The generation of these constrained density fields is an application and elaboration of the Hoffman & Ribak (1991) prescription. They showed that there is a simple and elegant solution to achieve this if the constraints are linear functionals of the field. We have presented the implementation of our method following a comprehensive discussion of the fundamentals underlying their method.

A maximum of 21 characteristics is used to specify the density and velocity at and around the position of the peak. They can be divided into three groups:

[1] The scale and position of the peak. We identify a peak as a local maximum in the density field that has been smoothed by a Gaussian filter function with a characteristic scale R_G , although the formalism can be very easily extended to other filter functions. The peak may be positioned at an arbitrary position within the simulation box.

[2] The local density field. In total 10 constraints are needed to fully specify the density field in the immediate vicinity of the peak. The first one concerns the height of the peak. In addition, three constraints are needed to assure that the three first derivatives of the smooth density field vanish at its summit. Finally, the six second order derivatives of the smooth density field are set by specifying the compactness $\nabla^2 f_G$, the axis ratios and the orientation of the peak.

[3] The local gravitational field. The specification of the gravitational field around the peak introduces 8 additional constraints: the three components of the smoothed peculiar gravity at the location of the peak and the five independent components of the traceless tidal tensor. The resulting density field is sculpted in such a way that it induces the desired amount of net gravitational and tidal forces. We usually restrict ourselves to the growing mode component of the density field. In the linear clustering regime the peculiar gravity and tidal field are therefore directly proportional to the peculiar velocity and the shear, so that we commonly use the latter to specify the gravitational field constraints.

It may be worthwhile to point out that in a linear density fluctuation field several of the above quantities are correlated. For example, we find that there is a strong correlation between the tidal field tensor and the mass tensor, expressing itself in the tendency of the tidal field to align itself along the principal axes of the mass tensor.

The constraints that we consider here are linear functionals of the density fluctuation field f , and therefore can be written as convolutions of the field with a specific function. Consequently, it is most convenient to perform the relevant calculations in Fourier space. The generation of a constrained field realization basically consists of the sum of an arbitrary field realization with the convolution of the power spectrum with a function that is the weighted sum of the different constraint kernels, the weights depending on the specified values of the constraints and the values of the constraints for the unconstrained field (see eq. 26). The expressions for these constraint kernels are derived from the particular constraints to which they are related. In Appendix F we list the kernels used in our code.

The Hoffman-Ribak algorithm that we have described here is considerably faster and more generally applicable than the original Bertschinger (1987) algorithm. Its superior speed is due to the direct and simple way of sampling the residual field, rendering an iterative “simulated annealing” technique superfluous. Moreover, because it is a direct method it has the additional advantage of superior accuracy. Extensive testing of constrained field realizations showed that the implementation is very precise, leading to accuracies in the order of 0.01% for the imposed quantities. In the computer implementation of our code the constrained field is evaluated on a periodic three-dimensional lattice. This has the advantage of being able to perform the Fourier transforms by means of a Fast Fourier Transform, with the advantage of being considerably faster than methods based on a direct Fourier transform. A disadvantage of the FFT is that they have

a rather weak sampling at low k , while direct Fourier transforms enable a far better sampling in that range. In their multipole constrained field method Bond and Meyers (1993) therefore resort to direct Fourier transforms, resulting in an excellent sampling at low and intermediate k .

In addition to the fact that the Hoffman-Ribak method provides us with a fast, efficient and accurate method to generate constrained random fields it has two other important advantages. The first one is that the implementation of a large variety of constraints is relatively straightforward through the convolution integrals in Fourier space. Secondly, unlike most other efficient algorithms it is equally suitable and efficient for local and non-local constraints. Although the illustrations of the peak constraints in section 5 were mainly local in character, centered on one peak, the developed formalism allows the generation of numerous peaks and dips at different positions (see figure 1).

In our application to peaks we followed the philosophy that each of the constraints corresponds to a different physical quantity. Another class of possible applications of the Hoffman-Ribak procedure is the reconstruction of (linear) density fields from the measurement of the same physical quantity at several different positions inside a certain volume. A nice illustration of this is the work by Ganon & Hoffman (1993), who reconstructed the density field in the “local” universe from the observed velocity field sampled at 181 different positions within a sphere of $40h^{-1}$ Mpc around us, assuming that it is a realization of a standard cold dark matter field. They showed that the method recovers the main features of POTENT’s density field (Dekel, Bertschinger & Faber 1990), in particular the Great Attractor region. The interesting feature of this reconstruction application is that it creates high-resolution fields subject to the low-resolution data, for the given underlying model. It therefore offers the charming and interesting opportunity to set up initial conditions for N -body simulations from observations of the local Universe, so that the nonlinear evolution of our “local” Universe in a particular cosmological scenario can be studied. A related and promising application would be the construction of high-resolution microwave background maps from the large-scale anisotropies measured by COBE (Bunn et al. 1994).

This class of constraint problems, where the constraints consist of the value of the same physical quantity $\psi(\mathbf{r})$ at many different positions, offers the advantage that for every constraint the constraint-field correlation function $\xi_i(\mathbf{r}) = \langle \psi(\mathbf{r}_i) f(\mathbf{r}) \rangle \equiv \Upsilon(\mathbf{r} - \mathbf{r}_i)$ (see eq. 24) can be evaluated from the same general correlation function $\Upsilon(\mathbf{x})$. The same is true for the constraint-constraint correlation function ξ_{ij} . In particular, this will be a great advantage if the constraint values are imposed at equally spaced points on a grid. This is the approach followed by Ganon & Hoffman (1993), who determined the constraint values for the velocity potential on a grid by spatial interpolation from observed values of the peculiar velocity. The computation of the required values of $\xi_i(\mathbf{x})$ and the inverse constraint-constraint correlation matrix ξ_{ij}^{-1} can then be simply accomplished by two FFTs. This can be easily seen from the following. Because the quantity $\psi(\mathbf{x})$ is a linear functional of the density field $f(\mathbf{x})$, its Fourier transform $\hat{\psi}(\mathbf{k})$ is a product of the Fourier transform $\hat{f}(\mathbf{k})$ of the field $f(\mathbf{x})$ with a kernel function $\hat{h}(\mathbf{k})$, $\hat{\psi}(\mathbf{k}) = \hat{h}(\mathbf{k})\hat{f}(\mathbf{k})$. Examples of such fields $\psi(\mathbf{x})$ are the gravitational potential, the peculiar velocity in the linear regime, or the temperature variations in the cosmic background radiation field. After evaluating the corresponding expressions for $\hat{h}(\mathbf{k})$ at wavenumbers \mathbf{k}_p (compare the kernel functions listed in Appendix F), the

values of $\xi_i(\mathbf{x}_j) = \langle \psi(\mathbf{x}_i) f(\mathbf{x}_j) \rangle$ and $\xi_{ij} = \langle \psi(\mathbf{x}_i) \psi(\mathbf{x}_j) \rangle$ can be found from $\xi_i(\mathbf{x}_j) = \Upsilon(\mathbf{x}_i - \mathbf{x}_j)$ and $\xi_{ij} = \Psi(\mathbf{x}_i - \mathbf{x}_j)$, where

$$\begin{aligned}\Upsilon(\mathbf{x}) &= \frac{1}{N} \sum_{p=0}^{N-1} \hat{h}(\mathbf{k}_p) P(k_p) e^{-i\mathbf{k}_p \cdot \mathbf{x}} \\ \Psi(\mathbf{x}) &= \frac{1}{N} \sum_{p=0}^{N-1} \hat{h}(\mathbf{k}_p)^2 P(k_p) e^{-i\mathbf{k}_p \cdot \mathbf{x}}.\end{aligned}\tag{113}$$

In fact, the inverse matrix of ξ_{ij} can be found directly and very simply from $\xi_{kl}^{-1} = \Theta(\mathbf{x}_k - \mathbf{x}_l)$, where $\Theta(\mathbf{x})$ is the inverse of $\Psi(\mathbf{x})$, i.e. $\Psi(\mathbf{x}_k - \mathbf{x}_i) \Theta(\mathbf{x}_i - \mathbf{x}_l) = \delta_{kl}$, and therefore given by the Fourier sum

$$\Theta(\mathbf{x}) = \frac{1}{N} \sum_{p=0}^{N-1} \frac{1}{\hat{h}(\mathbf{k}_p)^2 P(k_p)} e^{-i\mathbf{k}_p \cdot \mathbf{x}}.\tag{114}$$

The computation of the discrete Fourier sums $\Upsilon(\mathbf{x})$ and $\Theta(\mathbf{x})$ is accomplished by a FFT, so that the computational cost is only $\mathcal{O}(N \log N)$. Note that because of the periodic boundary conditions intrinsic to the FFT each coordinate of \mathbf{x}_i can only attain half of the values along each axis, so that in total only $\frac{1}{8}$ of the computational box is used for the field reconstruction. Finally, the independent Fourier components of the unconstrained field $\tilde{f}(\mathbf{x})$ are generated. The subsequent computation of $\tilde{f}(\mathbf{x})$ itself demands one FFT, and the computation of the corresponding constraint values \tilde{c}_j involves another FFT (compare eq. 38). Combining all these results in the final evaluation of the constrained field according to equation (35) consists of the computation of the double product $\xi_i(\mathbf{x}) \xi_{ij}^{-1} (c_j - \tilde{c}_j)$ for every point \mathbf{x}_j , making it an $\mathcal{O}(N^3)$ formalism. However, unlike the formalism developed in section 3, this procedure does not involve a very costly matrix inversion of ξ_{ij} , implying it to be far more efficient and the method of choice for this particular class of applications. On the other hand, when each of the M constraint quantities have a different character, concern different scales, arbitrary non-grid positions, or different filters, this procedure cannot be straightforwardly applied. In those cases a formalism similar to the one presented in this paper is automatically implied.

As a final note we should issue a cautionary remark on the practical implementation of our constrained random field code. The initial density fields are set up in a box with periodic boundary conditions. This means that the mean density of the box is exactly equal to the mean density of the Universe. The structure generated within the box is therefore not entirely typical, since overdense regions must necessarily be surrounded by low-density regions. This need not be true in general, from the theory of Gaussian random field we know that peaks tend to cluster. The simulation box should therefore not be taken too small, the resulting structure might be very atypical. Evidently, this conflicts with the demand to make the box as small as possible to achieve the highest possible resolution. The chosen box size should therefore be a compromise between these two.

In summary, we can conclude that the Hoffman-Ribak method provides a powerful and elegant tool to study the formation and evolution of specific cosmological objects in great detail under ideal conditions. The tools developed in this paper should essentially be regarded to constitute a

laboratory equipment set*. They allow us to set up very specific conditions for the objects under study. A sequence of experiments based on a range of different circumstances will subsequently yield a maximum of insight into the systematic dependence of structure formation on specific physical quantities. By concentrating on one specific application, peaks in the density field, we hope to have provided a recipe for constructing similar applications and extensions for different quantities in fields of a possibly different character. A straightforward extension of our formalism will for example be to consider peaks in the gravitational potential field instead of in the density field.

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* our FORTRAN computer code will be available upon request, and will be incorporated as part of the COSMICS package, Bertschinger 1995

Appendix A. The intersection of a sphere and a polygon

In section 2.2, equation 19, we saw that imposing the set of constraints $\Gamma = \{C_i[f; \mathbf{x}_i] = c_i; i = 1, \dots, M\}$ is equivalent to a change of the action $S[f]$ into

$$2S[f] = \int \int f(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) f(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2 - C^t \xi_{ij}^{-1} C, \quad (\text{A1})$$

with ξ_{ij} the $(ij)^{th}$ element of the matrix $\mathbf{Q} = \langle C^t C \rangle$.

The i^{th} constraint $C_i(\mathbf{x}_i)$ (in this appendix we will use the simplifying notation $C_i(\mathbf{x})$ for $C_i[f; \mathbf{x}]$) can be written as a convolution with a Dirac delta function $\delta_D(\mathbf{x})$,

$$C_i(\mathbf{x}_i) = \int d\mathbf{x}_2 \delta_D(\mathbf{x}_2) C_i(\mathbf{x}_i - \mathbf{x}_2) = \int d\mathbf{x}_1 \int d\mathbf{x}_2 \xi(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) C_i(\mathbf{x}_i - \mathbf{x}_2), \quad (\text{A2})$$

where we have used the fact that $K(\mathbf{x})$ is the functional inverse of the correlation function $\xi(\mathbf{x})$ (eq. 7, section 2.1). By using the convolution theorem we can express this double convolution integral in Fourier space as

$$C_i(\mathbf{x}_i) = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{C}_i(\mathbf{k}) P(\mathbf{k}) \hat{K}(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}_i}, \quad (\text{A3})$$

where $\hat{C}_i(\mathbf{k})$ is the Fourier transform of $C_i(\mathbf{x})$,

$$C_i(\mathbf{x}) = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{C}_i(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}}, \quad (\text{A4})$$

and $P(\mathbf{k}) = P(k)$, the spectral density, and $\hat{K}(\mathbf{k})$ the Fourier transforms of $\xi(\mathbf{x})$ and $K(\mathbf{x})$ respectively (eq. B5). The formal definition for the spectral density $P(k)$ is (Bertschinger 1992)

$$(2\pi)^3 P(k_1) \delta_D(\mathbf{k}_1 - \mathbf{k}_2) = \langle \hat{f}(\mathbf{k}_1) \hat{f}^*(\mathbf{k}_2) \rangle, \quad (\text{A5})$$

with $\delta_D(\mathbf{k}_1 - \mathbf{k}_2)$ the Dirac delta function. In an analogous fashion a function $\hat{P}_i(\mathbf{k})$ can be introduced,

$$(2\pi)^3 \hat{P}_i(\mathbf{k}_1) \delta_D(\mathbf{k}_1 - \mathbf{k}_2) = \langle \hat{C}_i(\mathbf{k}_1) \hat{f}^*(\mathbf{k}_2) \rangle, \quad (\text{A6})$$

from which we obtain, in combination with equation (A5), a relation between $\hat{f}(\mathbf{k})$ and $\hat{C}_i(\mathbf{k})$,

$$\frac{\langle \hat{f}(\mathbf{k}_1) \hat{f}^*(\mathbf{k}_2) \rangle}{P(k_1)} = \frac{\langle \hat{C}_i(\mathbf{k}_1) \hat{f}^*(\mathbf{k}_2) \rangle}{\hat{P}_i(\mathbf{k}_1)} \Rightarrow \hat{C}_i(\mathbf{k}) = \frac{\hat{P}_i(\mathbf{k})}{P(k)} \hat{f}(\mathbf{k}). \quad (\text{A7})$$

By subsequently inserting this relation in the Fourier integral of equation (A3), and using definition (A6), we get

$$\begin{aligned}
C_i(\mathbf{x}_i) &= \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{f}(\mathbf{k}) \hat{P}_i(\mathbf{k}) \hat{K}(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}_i} \\
&= \int \int \frac{d\mathbf{k}_1}{(2\pi)^3} \frac{d\mathbf{k}_2}{(2\pi)^3} \hat{f}(\mathbf{k}_1) \hat{K}(\mathbf{k}_1) \langle \hat{C}_i(\mathbf{k}_1) \hat{f}^*(\mathbf{k}_2) \rangle e^{-i\mathbf{k}_1 \cdot \mathbf{x}_i}, \\
&= \int \int f(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \xi_i(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2
\end{aligned} \tag{A8}$$

where the function $\xi_i(\mathbf{x})$ is the field-constraint correlation function and the Fourier transform of $\hat{P}_i(\mathbf{k})$,

$$\xi_i(\mathbf{x}) \equiv \langle f(\mathbf{x}) C_i(\mathbf{x}_i) \rangle = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{P}_i(\mathbf{k}) e^{-i\mathbf{k} \cdot (\mathbf{x}_i - \mathbf{x})}. \tag{A9}$$

Since the field $f(\mathbf{x})$ also obeys the constraints $C_j = c_j$ the expression $C_i(\mathbf{x}_i) \xi_{ij}^{-1} C_j(\mathbf{x}_j)$ in equation (A1) can be replaced by

$$C_i \xi_{ij}^{-1} C_j = \int \int f(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \bar{f}(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2, \tag{A10}$$

where we have defined the field $\bar{f}(\mathbf{x})$ by

$$\bar{f}(\mathbf{x}) \equiv \xi_i(\mathbf{x}) \xi_{ij}^{-1} c_j. \tag{A11}$$

The constrained action $S[f]$ in equation (A1) can therefore be written as

$$\begin{aligned}
2S[f] &= \int \int f(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \{f(\mathbf{x}_2) - \bar{f}(\mathbf{x}_2)\} d\mathbf{x}_1 d\mathbf{x}_2 \\
&= \int \int \{f(\mathbf{x}_1) - \bar{f}(\mathbf{x}_1)\} K(\mathbf{x}_1 - \mathbf{x}_2) \{f(\mathbf{x}_2) - \bar{f}(\mathbf{x}_2)\} d\mathbf{x}_1 d\mathbf{x}_2 + \\
&\quad \int \int \bar{f}(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \{f(\mathbf{x}_2) - \bar{f}(\mathbf{x}_2)\} d\mathbf{x}_1 d\mathbf{x}_2.
\end{aligned} \tag{A12}$$

It can be easily shown that the second term on the right hand side of equation (A12) is equal to zero because

$$\int \int \bar{f}(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \bar{f}(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2 = \int \int f(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \bar{f}(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2 = c_i \xi_{ij}^{-1} c_j. \tag{A13}$$

This relation follows directly from equation (A10) for the second integral, while for the first integral it follows from the fact that

$$\begin{aligned}
\int \int \bar{f}(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \bar{f}(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2 &= \xi_{ij}^{-1} \xi_{kl}^{-1} c_j c_l \int \int \xi_i(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \xi_k(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2 \\
&= \xi_{ij}^{-1} \xi_{kl}^{-1} \xi_{ki} c_j c_l = c_l \xi_{lj}^{-1} c_j
\end{aligned} \tag{A14}$$

where we have used the substitution of the Fourier expression of $\xi_i(\mathbf{x})$ (eq. A9) to evaluate the integral,

$$\begin{aligned}
\int \int \xi_i(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) \xi_k(\mathbf{x}_2) d\mathbf{x}_1 d\mathbf{x}_2 &= \\
&= \int \int \frac{d\mathbf{k}_1}{(2\pi)^3} \frac{d\mathbf{k}_2}{(2\pi)^3} \hat{P}_i(\mathbf{k}_1) \hat{K}(\mathbf{k}_2) \hat{P}_k^*(\mathbf{k}_2) (2\pi)^3 \delta_D(\mathbf{k}_2 - \mathbf{k}_1) e^{-i\mathbf{k}_1 \cdot \mathbf{x}_i} e^{i\mathbf{k}_2 \cdot \mathbf{x}_k} \\
&= \int \int \frac{d\mathbf{k}_1}{(2\pi)^3} \frac{d\mathbf{k}_2}{(2\pi)^3} \langle \hat{C}_k^*(\mathbf{k}_2) \hat{C}_i(\mathbf{k}_1) \rangle e^{-i\mathbf{k}_1 \cdot \mathbf{x}_i} e^{i\mathbf{k}_2 \cdot \mathbf{x}_k} = \langle C_k(\mathbf{x}_k) C_i(\mathbf{x}_i) \rangle = \xi_{ki}.
\end{aligned} \tag{A15}$$

The transition from the 2nd to 3rd line in equation (A15) has been made by combining equation (A6) and (A7),

$$(2\pi)^3 \hat{P}_i(\mathbf{k}_1) \delta_D(\mathbf{k}_1 - \mathbf{k}_2) = \frac{P(\mathbf{k}_2)}{\hat{P}_k^*(\mathbf{k}_2)} \langle \hat{C}_k^*(\mathbf{k}_2) \hat{C}_i(\mathbf{k}_1) \rangle, \tag{A16}$$

and the fact that $P(k) = 1/\hat{K}(\mathbf{k})$ (see eq. B6, app. B).

By defining the “residual field” $F(\mathbf{x}) = f(\mathbf{x}) - \bar{f}(\mathbf{x})$ we can therefore conclude from equation (A12) that the constrained action $S[f] = S[F]$ can be written as

$$2S[F] = \int d\mathbf{x}_1 \int d\mathbf{x}_2 F(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) F(\mathbf{x}_2), \tag{A17}$$

which is the expression needed in Section 2.2.

Appendix B: Diagonalisation of the action $S[F]$

In this appendix we will rewrite the action $S[F]$ (eq. 20),

$$S[F] = \frac{1}{2} \int d\mathbf{x}_1 \int d\mathbf{x}_2 F^*(\mathbf{x}_1) K(\mathbf{x}_1 - \mathbf{x}_2) F(\mathbf{x}_2). \tag{B1}$$

in terms of the Fourier transform $\hat{F}(\mathbf{k})$ of the fluctuation field $F(\mathbf{x})$,

$$F(\mathbf{x}) = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{F}(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}}. \tag{B2}$$

The kernel $K(\mathbf{x})$ in equation (B1) is the functional inverse of the correlation function $\xi(x)$ (eq. 7),

$$\int d\mathbf{x} K(\mathbf{x}_1 - \mathbf{x}) \xi(\mathbf{x} - \mathbf{x}_2) = \delta_D(\mathbf{x}_1 - \mathbf{x}_2). \tag{B3}$$

By virtue of the convolution theorem this equation is equivalent to

$$\int \frac{d\mathbf{k}}{(2\pi)^3} \hat{K}(\mathbf{k}) P(k) e^{i\mathbf{k} \cdot (\mathbf{x}_1 - \mathbf{x}_2)} = \delta_D(\mathbf{x}_1 - \mathbf{x}_2), \tag{B4}$$

where $\hat{K}(\mathbf{k})$ and $P(k)$ are the Fourier transform of $K(\mathbf{x})$ and $\xi(\mathbf{x})$,

$$K(\mathbf{x}) = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{K}(\mathbf{k}) e^{-i\mathbf{k} \cdot \mathbf{x}}, \quad \xi(\mathbf{x}) = \int \frac{d\mathbf{k}}{(2\pi)^3} P(k) e^{-i\mathbf{k} \cdot \mathbf{x}}. \tag{B5}$$

The identification of the left part of equation (B4) with the Fourier integral expression of the Dirac delta function implies that $\hat{K}(\mathbf{k}) = 1/P(k)$. Consequently,

$$K(\mathbf{x}) = \int \frac{d\mathbf{k}}{(2\pi)^3} \frac{1}{P(k)} e^{-i\mathbf{k}\cdot\mathbf{x}}. \quad (\text{B6})$$

Likewise, the insertion of $\hat{K}(\mathbf{k}) = 1/P(k)$ into the double convolution of equation (B1),

$$2S[F] = \int \frac{d\mathbf{k}}{(2\pi)^3} \hat{F}^*(\mathbf{k}) \hat{K}(\mathbf{k}) \hat{F}(\mathbf{k}), \quad (\text{B7})$$

yields the Fourier expression for $S[F]$,

$$S[f] = \int \frac{d\mathbf{k}}{(2\pi)^3} \frac{|\hat{\mathbf{F}}(\mathbf{k})|^2}{2P(k)}, \quad (\text{B8})$$

which is equation 29 in section 3.

Appendix C: A heuristic proof that $\mathcal{P}[F|\Gamma]$ is independent of c_i .

A field $f(\mathbf{x})$ can be viewed as an N -dimensional vector (f_1, \dots, f_N) in N -dimensional “field” space, with $N \rightarrow \infty$. The fields $f(\mathbf{x})$ that obey the set of M constraints $\Gamma = \{C_i[f; \mathbf{x}_i] = c_i; i = 1, \dots, M\}$ define an $(N-M)$ -dimensional hypersurface in this N -dimensional space. For reasons of convenience this hypersurface will also be denoted as Γ . The only restriction that we impose on the constraints C_i is that they are linear,

$$C_i[f_1 + f_2; \mathbf{x}] = C_i[f_1; \mathbf{x}] + C_i[f_2; \mathbf{x}]. \quad (\text{C1})$$

Each of the hypersurfaces Γ contain a special point $\bar{f}(\mathbf{x})$, the mean of the fields satisfying the constraints Γ ,

$$\bar{f}(\mathbf{x}) = \langle f(\mathbf{x}) | \Gamma \rangle = \xi_i(\mathbf{x}) \xi_{ij}^{-1} c_j, \quad (\text{C2})$$

where $\xi_i(\mathbf{x})$ is the cross-correlation between the field and the i^{th} constraint $C_i[f; \mathbf{x}]$, and ξ_{ij} the correlation between the i^{th} and j^{th} constraints, $C_i[f]$ and $C_j[f]$ (notice that in this notation we stress the functional character of the constraints). Both ξ_{ij} and $\xi_i(\mathbf{x})$ are defined in equation (24) (Section 2.2). Each of the fields $f(\mathbf{x})$ in Γ have a corresponding residual field $F(\mathbf{x})$, defined as the difference between the field $f(\mathbf{x})$ and the mean field $\bar{f}(\mathbf{x})$ of Γ ,

$$F(\mathbf{x}) \equiv f(\mathbf{x}) - \bar{f}(\mathbf{x}). \quad (\text{C3})$$

Imagine two arbitrarily chosen constraint hypersurfaces, the first one corresponding to the constraint set $\Gamma_1 = \{C_i[f; \mathbf{x}_i] = c_{i,1}; i = 1, \dots, M\}$ and the other one to the set $\Gamma_2 = \{C_i[f; \mathbf{x}_i] = c_{i,2}; i = 1, \dots, M\}$. The mean fields of the sets Γ_1 and Γ_2 are \bar{f}_1 and \bar{f}_2 . Consider the translation of an arbitrary field $f_1(\mathbf{x}) \in \Gamma_1$ by a field $T_{2,1}(\mathbf{x})$ into a field $f_T(\mathbf{x})$,

$$f_T(\mathbf{x}) \equiv f_1(\mathbf{x}) + T_{2,1}(\mathbf{x}), \quad (\text{C4})$$

where the translation $T_{2,1}(\mathbf{x})$ is defined by

$$T_{2,1}(\mathbf{x}) \equiv \bar{f}_2(\mathbf{x}) - \bar{f}_1(\mathbf{x}) = \xi_i(\mathbf{x}) \xi_{ij}^{-1}(c_{j,2} - c_{j,1}). \quad (\text{C5})$$

This definition of $T_{2,1}$ immediately implies that the mean field $\bar{f}_1(\mathbf{x})$ of Γ_1 is transformed into the mean field $\bar{f}_2(\mathbf{x})$ of Γ_2 . From equations (C4) and (C5) and the linearity of the constraints C_i we can infer that

$$\begin{aligned} C_i[f_T] &= C_i[f_1] + C_i[T_{2,1}] \\ &= C_i[f_1] + C_i[\bar{f}_2] - C_i[\bar{f}_1] \\ &= c_{i,1} + c_{i,2} - c_{i,1} = c_{i,2}, \end{aligned} \quad (\text{C6})$$

The field $f_T(\mathbf{x})$ therefore obeys the constraint set Γ_2 . This is true regardless of the field $f_1(\mathbf{x}) \in \Gamma_1$. Moreover, the inverse translation $-T_{2,1}$ transforms the resulting field $f_2(\mathbf{x})$ back into $f_1(\mathbf{x})$. The two hypersurfaces Γ_1 and Γ_2 are therefore linked by a one-to-one mapping, so that

$$\mathcal{P}[f_1|\Gamma_1] = \mathcal{P}[f_2|\Gamma_2], \quad (\text{C7})$$

where $\mathcal{P}[f_1|\Gamma_1]$ is the probability of having a specific field $f_1(\mathbf{x})$ under the condition that they satisfy the constraints Γ_1 , and $f_2(\mathbf{x})$ is the field in the hypersurface Γ_2 that is linked to $f_1(\mathbf{x})$ by the translation $T_{2,1}(\mathbf{x})$ (eq. C4). The conditional probabilities for the corresponding residual fields $F_1(\mathbf{x}) \equiv (f_1(\mathbf{x}) - \bar{f}_1(\mathbf{x}))$ and $F_2(\mathbf{x}) \equiv (f_2(\mathbf{x}) - \bar{f}_2(\mathbf{x}))$ can be inferred from equation (C7),

$$\mathcal{P}[F_1|\Gamma_1] = \mathcal{P}[f_1|\Gamma_1] = \mathcal{P}[f_2|\Gamma_2] = \mathcal{P}[F_2|\Gamma_2]. \quad (\text{C8})$$

Finally, consider the transformation of the residual field $F_1(\mathbf{x})$ under the translation $T_{2,1}$,

$$\begin{aligned} F_1(\mathbf{x}) &\equiv f_1(\mathbf{x}) - \bar{f}_1(\mathbf{x}) \\ &= (f_1(\mathbf{x}) + T_{2,1}(\mathbf{x})) - (\bar{f}_1(\mathbf{x}) + T_{2,1}(\mathbf{x})) \\ &= f_2(\mathbf{x}) - \bar{f}_2(\mathbf{x}) = F_2(\mathbf{x}). \end{aligned} \quad (\text{C9})$$

In other words, the residual field $F(\mathbf{x})$ is invariant under the translation $T_{2,1}$, i.e. $F_1 = F = F_2$, which in combination with equation (C8) implies that

$$\mathcal{P}[F|\Gamma_1] = \mathcal{P}[F_1|\Gamma_1] = \mathcal{P}[F_2|\Gamma_2] = \mathcal{P}[F|\Gamma_2]. \quad (\text{C10})$$

This is the result that we intended to prove.

Appendix D: The variance $\langle F^2(\mathbf{x})|\Gamma \rangle$ of the residual field $F(\mathbf{x})$.

A derivation will be given for the expression for the variance $\langle F^2(\mathbf{x})|\Gamma \rangle$ of the residual field belonging to the constraint set Γ . The residual field $F(\mathbf{x})$ is the difference between a field $f(\mathbf{x})$ obeying the constraint set Γ and the mean $\bar{f}(\mathbf{x}) \equiv \langle F(\mathbf{x})|\Gamma \rangle$ of all these fields,

$$\bar{f}(\mathbf{x}) = \xi_i(\mathbf{x}) \xi_{ij}^{-1} c_j. \quad (\text{D1})$$

The crucial observation that $\mathcal{P}[F|\Gamma]$, the probability of having a residual field $F(\mathbf{x})$ satisfying a particular set of constraints Γ , is independent of the numerical value c_i of the constraints Γ ,

$$\mathcal{P}[F|\Gamma_1] = \mathcal{P}[F|\Gamma_2], \quad (\text{D2})$$

implies that

$$\langle F^2(\mathbf{x})|\Gamma \rangle = \langle F^2(\mathbf{x}) \rangle, \quad (\text{D3})$$

where $\langle F^2 \rangle$ is the variance in all possible realizations of the field, and $\langle F^2|\Gamma \rangle$ the variance for the ones that obey the constraint set Γ . From equation (D3) we find

$$\begin{aligned} \langle F^2(\mathbf{x}) \rangle &= \int \mathcal{P}[\Gamma] \langle F^2(\mathbf{x})|\Gamma \rangle = \int \mathcal{P}[\Gamma] \langle (f(\mathbf{x}) - \bar{f}(\mathbf{x}))^2|\Gamma \rangle \\ &= \int \mathcal{P}[\Gamma] \left\{ \langle f^2(\mathbf{x})|\Gamma \rangle - \langle f(\mathbf{x})|\Gamma \rangle^2 \right\}, \end{aligned} \quad (\text{D4})$$

where $\mathcal{P}[\Gamma]$ is the integrated probability of all realizations that obey the constraint set Γ . Evaluation of the first part of the integral in (D4) yields

$$\begin{aligned} \int \mathcal{P}[\Gamma] \langle f^2(\mathbf{x})|\Gamma \rangle &= \int \mathcal{P}[\Gamma] \int \mathcal{P}[f(\mathbf{x})|\Gamma] f^2(\mathbf{x}) = \int \mathcal{P}[\Gamma] \mathcal{P}[f(\mathbf{x})|\Gamma] f^2(\mathbf{x}) \\ &= \int \mathcal{P}[f(\mathbf{x})] f^2(\mathbf{x}) = \langle f^2(\mathbf{x}) \rangle = \sigma_0^2, \end{aligned} \quad (\text{D5})$$

where σ_0^2 is the general variance of the density field fluctuations. In the derivation of (D5) we have used the fact that $\mathcal{P}[f|\Gamma]$ is the product of the probability $\mathcal{P}[\Gamma]$ with the conditional probability of having the field $f(\mathbf{x})$ under the condition that it obeys Γ , $\mathcal{P}[f|\Gamma]$ (equation 13, section 2.2).

To evaluate the second part of the integral we use the expression for the mean field $\langle f|\Gamma \rangle$ in equation (D1),

$$\begin{aligned} \int \mathcal{P}[\Gamma] \langle f(\mathbf{x})|\Gamma \rangle^2 &= \int \mathcal{P}[\Gamma] \xi_i(\mathbf{x}) \xi_{ij}^{-1} c_j c_l \xi_{kl}^{-1} \xi_k(\mathbf{x}) \\ &= \xi_i(\mathbf{x}) \xi_{ij}^{-1} \left\{ \int \mathcal{P}(\Gamma) C_j(\mathbf{x}_j) C_l(\mathbf{x}_l) \right\} \xi_{lk}^{-1} \xi_k(\mathbf{x}) \\ &= \xi_i(\mathbf{x}) \xi_{ij}^{-1} \langle C_j C_l \rangle \xi_{lk}^{-1} \xi_k(\mathbf{x}) = \xi_i(\mathbf{x}) \xi_{ij}^{-1} \xi_{jl} \xi_{lk}^{-1} \xi_k(\mathbf{x}) = \xi_i(\mathbf{x}) \xi_{ik}^{-1} \xi_k(\mathbf{x}). \end{aligned} \quad (\text{D6})$$

By inserting equations (D5) and (D6) into equation (D4) and using equation (D3) we find

$$\langle F^2(\mathbf{x})|\Gamma \rangle = \sigma_0^2 - \xi_i(\mathbf{x}) \xi_{ij}^{-1} \xi_j(\mathbf{x}), \quad (\text{D7})$$

which is the intended expression.

Appendix E: Shape and orientation of a peak in a random field.

The second order Taylor expansion of a density field around a peak or dip at position \mathbf{x}_d in a density field $f(\mathbf{x})$ is given by equation (51), which we repeat here for convenience,

$$f_G(\mathbf{x}) = f_G(\mathbf{x}_d) + \frac{1}{2} \sum_{i,j=1}^3 \frac{\partial^2 f_G}{\partial x_i \partial x_j}(\mathbf{x}_d) (x_i - x_{d,i})(x_j - x_{d,j}). \quad (\text{E1})$$

This quadratic equation can be written in its canonical form by transforming to the coordinate system $\mathbf{x}' = \{x'_1, x'_2, x'_3\}$ whose axes are aligned along the eigenvectors of the matrix $\nabla_i \nabla_j f_G$. If the eigenvalues of $-\nabla_i \nabla_j f_g$ are λ_1, λ_2 and λ_3 , equation (E1) becomes

$$f_G(\mathbf{x}') = f_G(0) - \frac{1}{2} \sum_{i=1}^3 \lambda_i x_i'^2, \quad (\text{E2})$$

where we have chosen the origin of \mathbf{x}' to coincide with the position of the peak or dip. In the case of a peak the λ_i have a negative value, for a dip they have a positive value. From equation (E2) we see that the isodensity surface $f_G = F$ is a triaxial ellipsoid whose principal axes are oriented along the coordinate axes, with semiaxes given by

$$a_i = \left[\frac{2(\nu_d \sigma_0(R_G) - F)}{\lambda_i} \right]^{1/2}, \quad i = 1, \dots, 3. \quad (\text{E3})$$

In equation (E3) the central height $f_G(\mathbf{x}_d)$ of the overdensity is expressed in units of $\sigma_0(R_G)$, i.e. $f_G(\mathbf{x}_d) = \nu_c \sigma_0(R_G)$.

From equation (E3) and the fact that the shape of a triaxial ellipsoid is fully specified by its two axis ratios $a_{12} \equiv (a_1/a_2)$ and $a_{13} \equiv (a_1/a_3)$ we can infer that constraints on the shape of the overdensity result in constraints on the ratio of λ_i 's,

$$\left(\frac{\lambda_2}{\lambda_1} \right) = a_{12}^2, \quad \left(\frac{\lambda_3}{\lambda_1} \right) = a_{13}^2. \quad (\text{E4})$$

The actual magnitude of the λ_i 's depends on the steepness of the density profile around the peak. This steepness is specified by the Laplacian $\nabla^2 f_G$, as can be observed from the expansion of the density profile equation (E2) in spherical coordinates (x, θ, φ) ,

$$f_G(\mathbf{x}) = f_G(\mathbf{x}_d) + \nabla^2 f_G(\mathbf{x}_d) \frac{x^2}{2} \{1 + A(\theta, \varphi)\}. \quad (\text{E5})$$

$A(\theta, \varphi)$ is a function of the direction (θ, φ) that describes the asphericity of the peak via its dependence on the parameters λ_1, λ_2 and λ_3 . In deriving equation (E5) we used the relation between the λ_i and $\nabla^2 f_G$

$$\sum_{i=1}^3 \lambda_i = -\nabla^2 f_G(\mathbf{x}_d), \quad (\text{E6})$$

which can be obtained by double differentiation of equation (E2). Usually $\nabla^2 f_G$ is expressed in units of $\sigma_2(R_G) = \langle \nabla^2 f_G \nabla^2 f_G \rangle^{1/2}$, i.e. $\nabla^2 f_G(R_G) = -x_d \sigma_2(R_G)$. The expression for λ_1 is obtained by combination of equations (E4) and (E6),

$$\lambda_1 = \frac{x_d \sigma_2(R_G)}{(1 + a_{12}^2 + a_{13}^2)}. \quad (\text{E7})$$

Once the value of λ_1 has been determined, the values of λ_2 and λ_3 are obtained by multiplication of λ_1 by a_{12}^2 and a_{13}^2 respectively (equation E4).

The orientation of the peak with respect to the general coordinate system is described by the three Euler angles α , β and ψ (see Goldstein 1980). Here β is the angle between the smallest axis of the ellipsoid and the z -coordinate axis, α the angle between the line of nodes and the x -coordinate axis, and ψ the angle between the largest axis of the ellipsoid and the line of nodes. The line of nodes is the intersection of the xy -plane and the plane defined by the largest and second largest axis of the ellipsoid. The transformation matrix A_{ij} (equation 60, Sect. 4.2) is obtained from this definition of the Euler angles,

$$A = \begin{pmatrix} \cos \alpha \cos \psi - \cos \beta \sin \alpha \sin \psi & \sin \alpha \cos \psi + \cos \beta \cos \alpha \sin \psi & \sin \beta \sin \psi \\ -\cos \alpha \sin \psi - \cos \beta \sin \alpha \cos \psi & -\sin \alpha \sin \psi + \cos \beta \cos \alpha \cos \psi & -\sin \beta \cos \psi \\ \sin \beta \sin \alpha & -\sin \beta \cos \alpha & \cos \beta \end{pmatrix}. \quad (\text{E8})$$

This matrix describes the transformation from the coordinate system \mathbf{x}' defined by the principal axes of the ellipsoid to the general coordinate system \mathbf{x} ,

$$x'_i = \sum_{j=1}^3 A_{ij}(x_j - x_{d,j}), \quad i = 1, \dots, 3, \quad (\text{E9})$$

with \mathbf{x}_d the position of the centre of the peak. Thus, $x_i'^2$ transforms as

$$x_i'^2 = \sum_{j=1}^3 \sum_{k=1}^3 A_{ij} A_{ik} (x_j - x_{d,j})(x_k - x_{d,k}). \quad (\text{E10})$$

By inserting this transformation into the expression for the density profile (eq. E2) we obtain the following quadratic equation for the density profile in the general coordinate system \mathbf{x} ,

$$f_G(\mathbf{x}) = f_G(\mathbf{x}_d) - \frac{1}{2} \sum_{j,k=1}^3 \left\{ \sum_{i=1}^3 \lambda_i A_{ij} A_{ik} \right\} (x_j - x_{d,j})(x_k - x_{d,k}). \quad (\text{E11})$$

Because equation (E11) is equivalent to equation (E1) we obtain the following relationship between the second derivatives of f_G and the orientation, shape and steepness of the dip or peak in the field f_G ,

$$\frac{\partial^2 f_G}{\partial x_i \partial x_j} = - \sum_{k=1}^3 \lambda_k A_{ki} A_{kj}, \quad i, j = 1, 2, 3. \quad (\text{E12})$$

This is the expression that we use in section 4.2.

Appendix F: Peak constraint kernels and values.

Here we present the explicit expressions for the 18 peak constraints and the corresponding kernels $\hat{H}_l(\mathbf{k})$, defined in equation (37), to give an overview and summary of the results in this paper.

The filter function $\hat{W}(\mathbf{k})$ is taken to be the one corresponding to a Gaussian filter function with smoothing length R_G ,

$$\hat{W}(\mathbf{k}) = e^{-k^2 R_G^2/2}. \quad (\text{F1})$$

The peak constraints are presented in 5 groups. The first group consists of the peak height constraint $f_G(\mathbf{x}_d)$, the second one of the three constraints on the first derivative of the field, $\nabla f_G(\mathbf{x}_d)$, and the third one of the second derivatives $\nabla_i \nabla_j f_G(\mathbf{x}_d)$. In addition, the fourth group contains the constraints on the peculiar velocity of the peak, $\mathbf{v}_G(\mathbf{x}_d)$, while the fifth group corresponds to the constraints on the five components of the shear, $\sigma_{G,ij}(\mathbf{x}_d)$,

$$f_G(\mathbf{x}_d) = \nu \sigma_0(R_G)$$

$$\hat{H}_1(\mathbf{k}) = e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial f_G}{\partial x_1}(\mathbf{x}_d) = 0$$

$$\hat{H}_2(\mathbf{k}) = ik_1 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial f_G}{\partial x_2}(\mathbf{x}_d) = 0$$

$$\hat{H}_3(\mathbf{k}) = ik_2 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial f_G}{\partial x_3}(\mathbf{x}_d) = 0$$

$$\hat{H}_4(\mathbf{k}) = ik_3 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial^2 f_G}{\partial x_1^2}(\mathbf{x}_d) = -\sum_{k=1}^3 \lambda_k A_{k1} A_{k1}$$

$$\hat{H}_5(\mathbf{k}) = -k_1^2 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial^2 f_G}{\partial x_2^2}(\mathbf{x}_d) = -\sum_{k=1}^3 \lambda_k A_{k2} A_{k2}$$

$$\hat{H}_6(\mathbf{k}) = -k_2^2 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial^2 f_G}{\partial x_3^2}(\mathbf{x}_d) = -\sum_{k=1}^3 \lambda_k A_{k3} A_{k3}$$

$$\hat{H}_7(\mathbf{k}) = -k_3^2 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial^2 f_G}{\partial x_1 \partial x_2}(\mathbf{x}_d) = -\sum_{k=1}^3 \lambda_k A_{k1} A_{k2}$$

$$\hat{H}_8(\mathbf{k}) = -k_1 k_2 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial^2 f_G}{\partial x_1 \partial x_3}(\mathbf{x}_d) = -\sum_{k=1}^3 \lambda_k A_{k1} A_{k3}$$

$$\hat{H}_9(\mathbf{k}) = -k_1 k_3 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$\frac{\partial^2 f_G}{\partial x_2 \partial x_3}(\mathbf{x}_d) = -\sum_{k=1}^3 \lambda_k A_{k2} A_{k3}$$

$$\hat{H}_{10}(\mathbf{k}) = -k_2 k_3 e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$g_{G,1}(\mathbf{x}_d) = \tilde{g}_1 \sigma_{g,pk}(R_G)$$

$$\hat{H}_{11}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \frac{ik_1}{k^2} e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$g_{G,2}(\mathbf{x}_d) = \tilde{g}_2 \sigma_{g,pk}(R_G)$$

$$\hat{H}_{12}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \frac{ik_2}{k^2} e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$g_{G,3}(\mathbf{x}_d) = \tilde{g}_3 \sigma_{g,pk}(R_G)$$

$$\hat{H}_{13}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \frac{ik_3}{k^2} e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$E_{G,11}(\mathbf{x}_d) = \tilde{\epsilon} \sigma_E(R_G) \sum_{k=1}^3 \mathcal{Q}(\varpi) T_{k1} T_{k1}$$

$$\hat{H}_{14}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \left(\frac{k_1^2}{k^2} - \frac{1}{3} \right) e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$E_{G,22}(\mathbf{x}_d) = \tilde{\epsilon} \sigma_E(R_G) \sum_{k=1}^3 \mathcal{Q}(\varpi) T_{k2} T_{k2}$$

$$\hat{H}_{15}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \left(\frac{k_2^2}{k^2} - \frac{1}{3} \right) e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$E_{G,12}(\mathbf{x}_d) = \tilde{\epsilon} \sigma_E(R_G) \sum_{k=1}^3 \mathcal{Q}(\varpi) T_{k1} T_{k2}$$

$$\hat{H}_{16}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \left(\frac{k_1 k_2}{k^2} \right) e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$E_{G,13}(\mathbf{x}_d) = \tilde{\epsilon} \sigma_E(R_G) \sum_{k=1}^3 \mathcal{Q}(\varpi) T_{k1} T_{k3}$$

$$\hat{H}_{17}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \left(\frac{k_1 k_3}{k^2} \right) e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$

$$E_{G,23}(\mathbf{x}_d) = \tilde{\epsilon} \sigma_E(R_G) \sum_{k=1}^3 \mathcal{Q}(\varpi) T_{k2} T_{k3}$$

$$\hat{H}_{18}(\mathbf{k}) = \frac{3}{2} \Omega H^2 \left(\frac{k_2 k_3}{k^2} \right) e^{-k^2 R_G^2/2} e^{i\mathbf{k} \cdot \mathbf{x}_d}$$